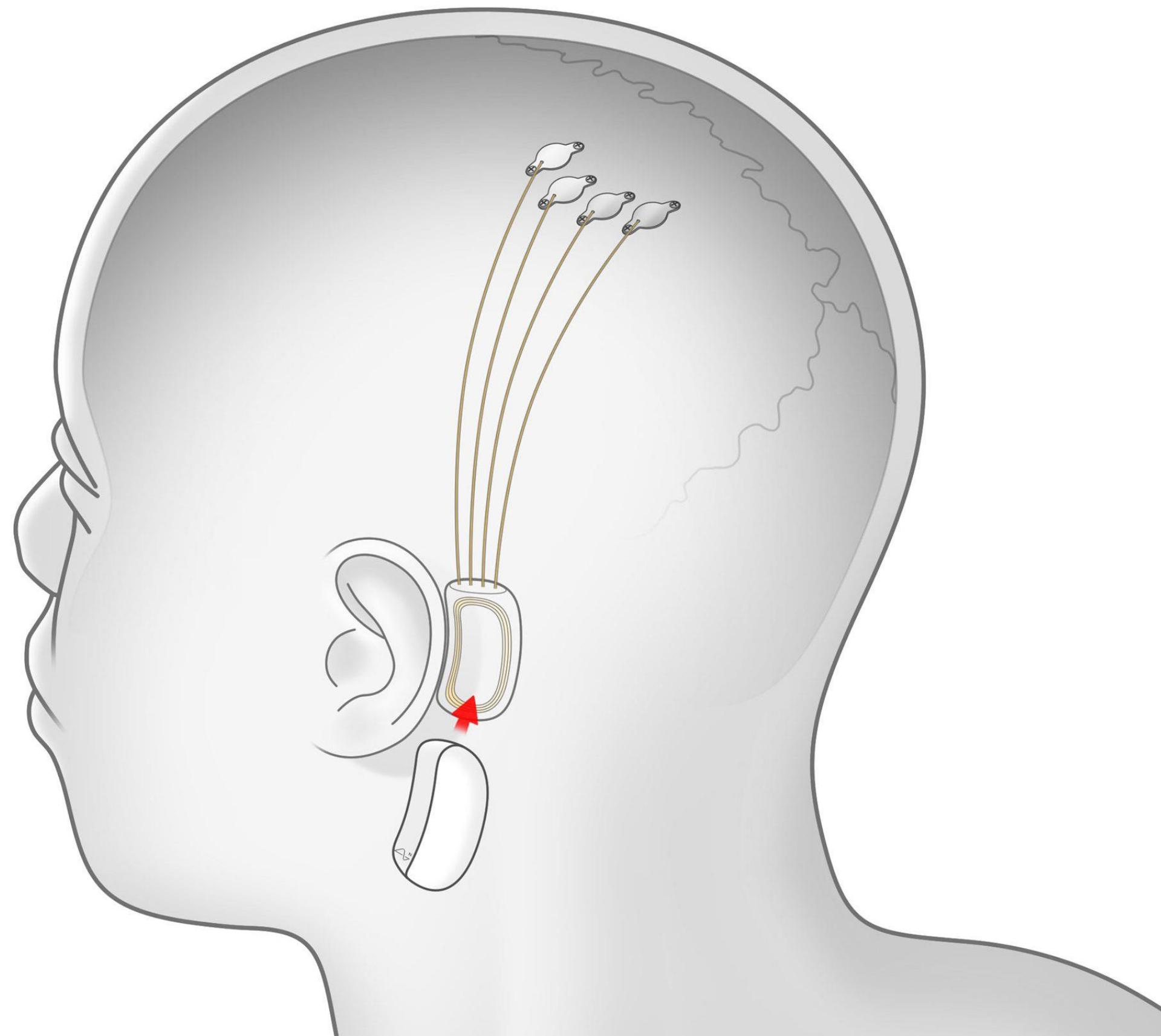


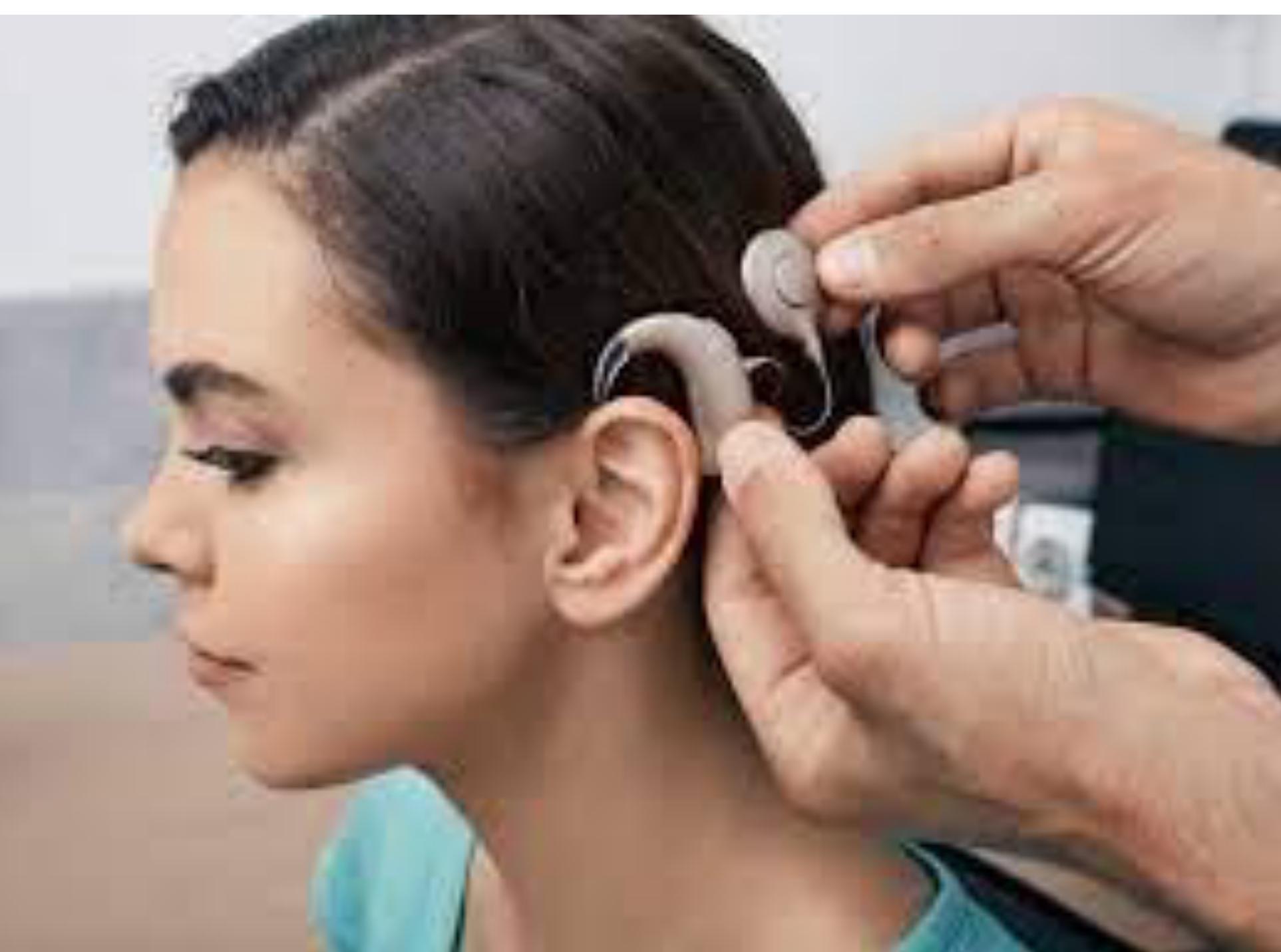
Bionic artificial limbs (2nd part)

Translational neuroengineering

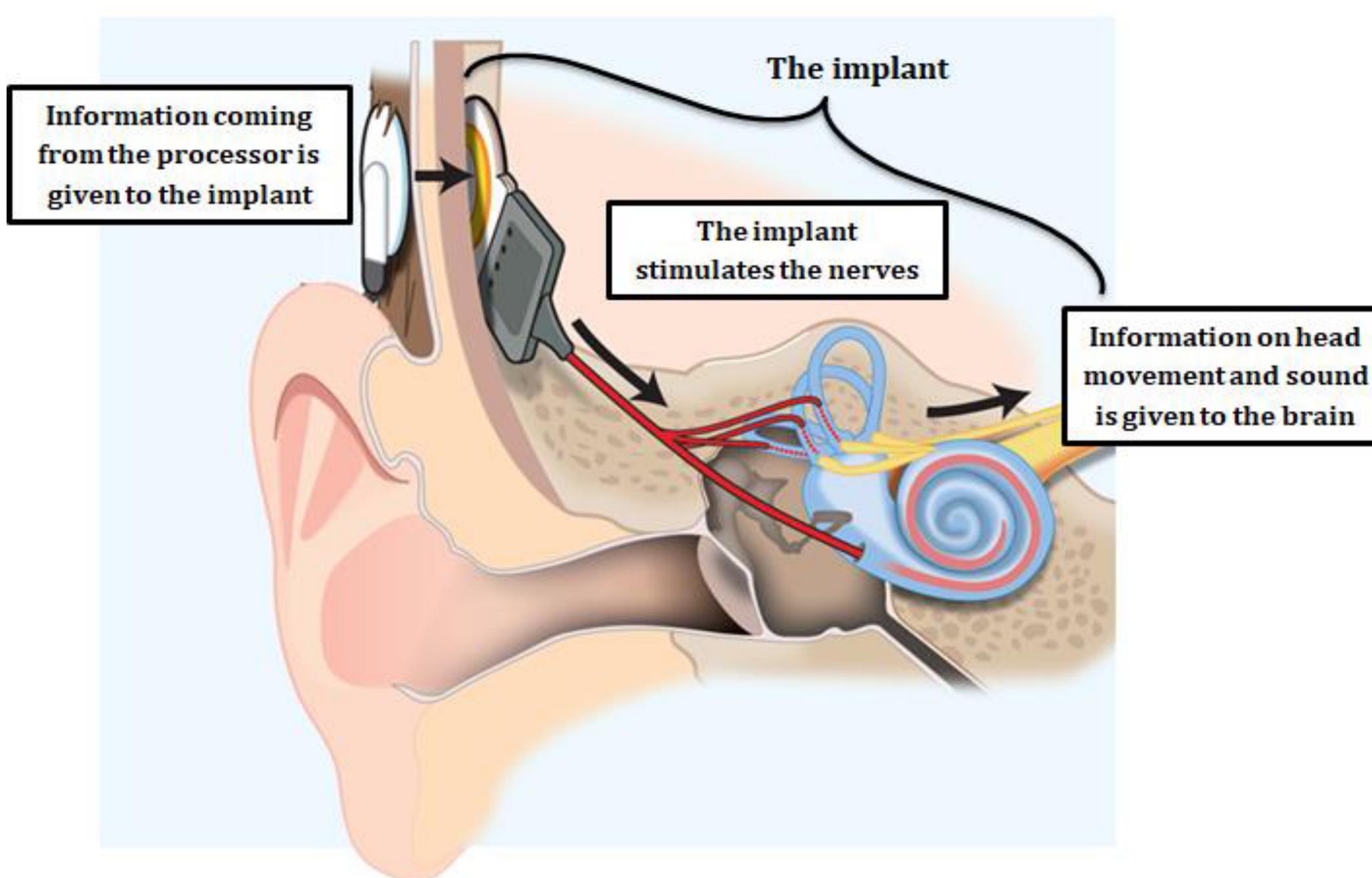
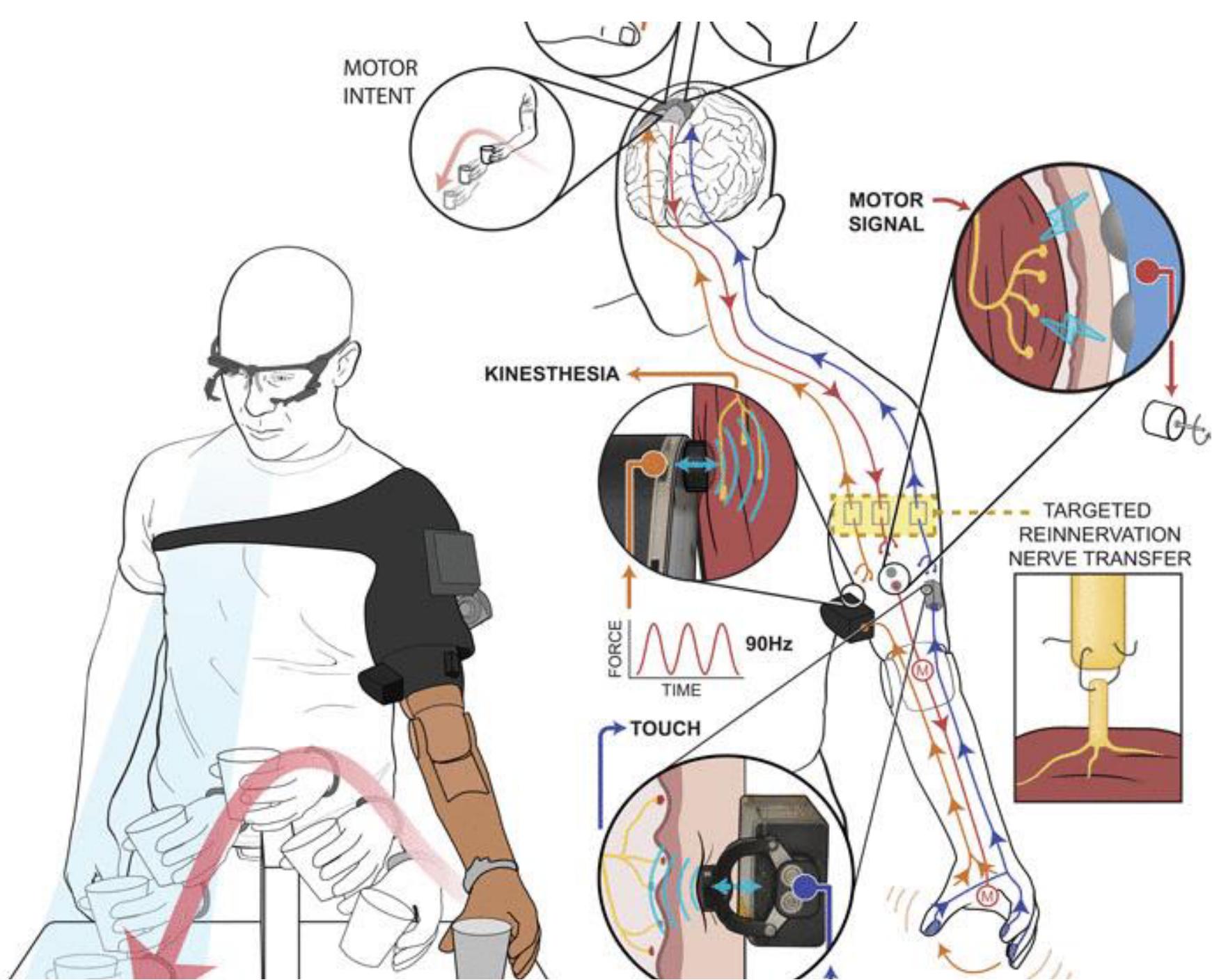
Silvestro Micera,
silvestro.micera@epfl.ch

Bertarelli Foundation Chair in Translational
Neuroengineering,
Center for Neuroprosthetics, EPFL, Geneva





■ Fundamentals of neuroengineering

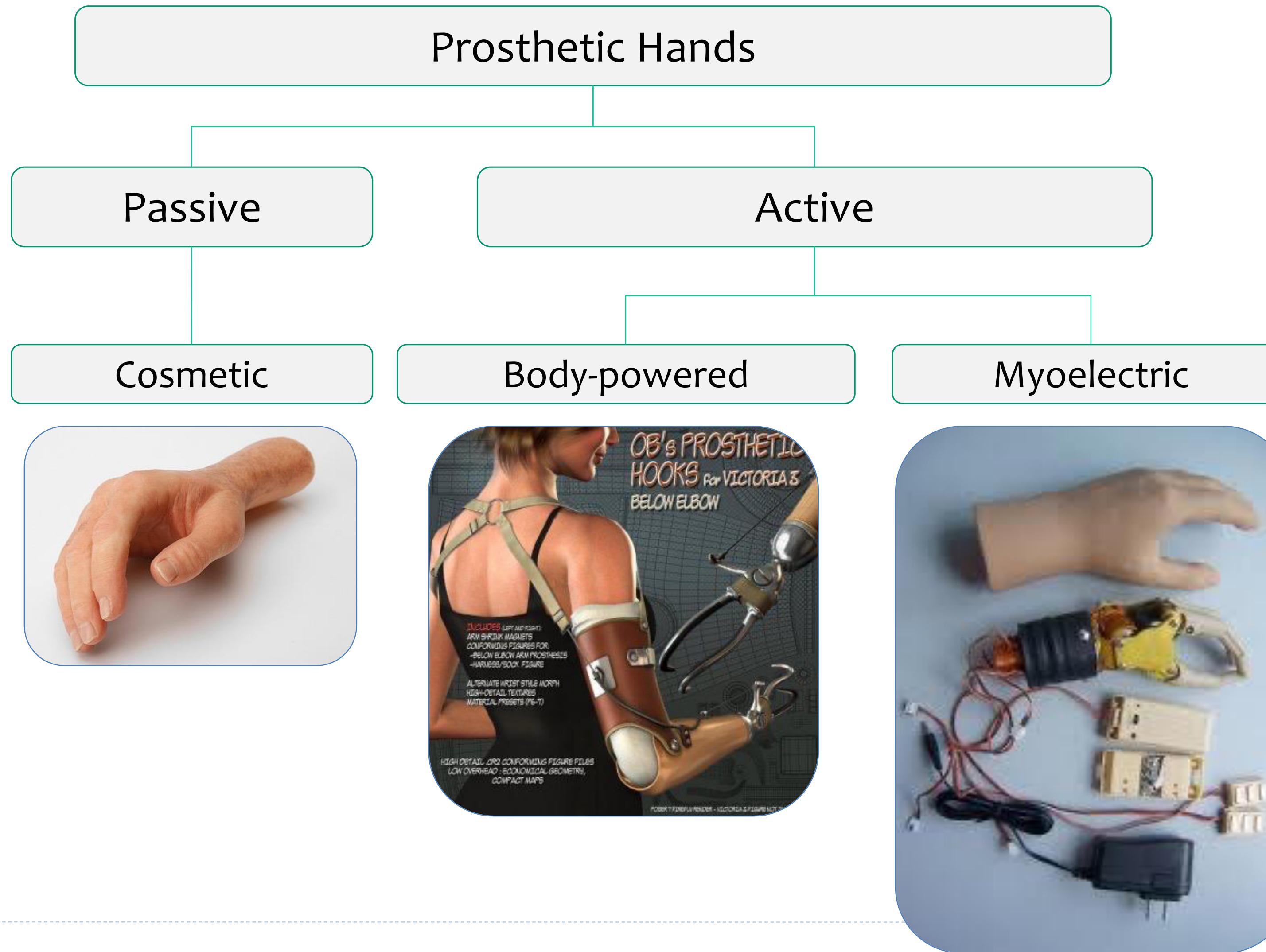


Bionic artificial hand



What can an amputee get today?

Hand Prosthesis



Mechatronic Design issues: adaptability

Problem: It's an hard task to **design, actuate, and control** a self-contained artificial hand with a number of degrees of freedom (DoF) equal or close to those in the biological human hand!

22 muscles

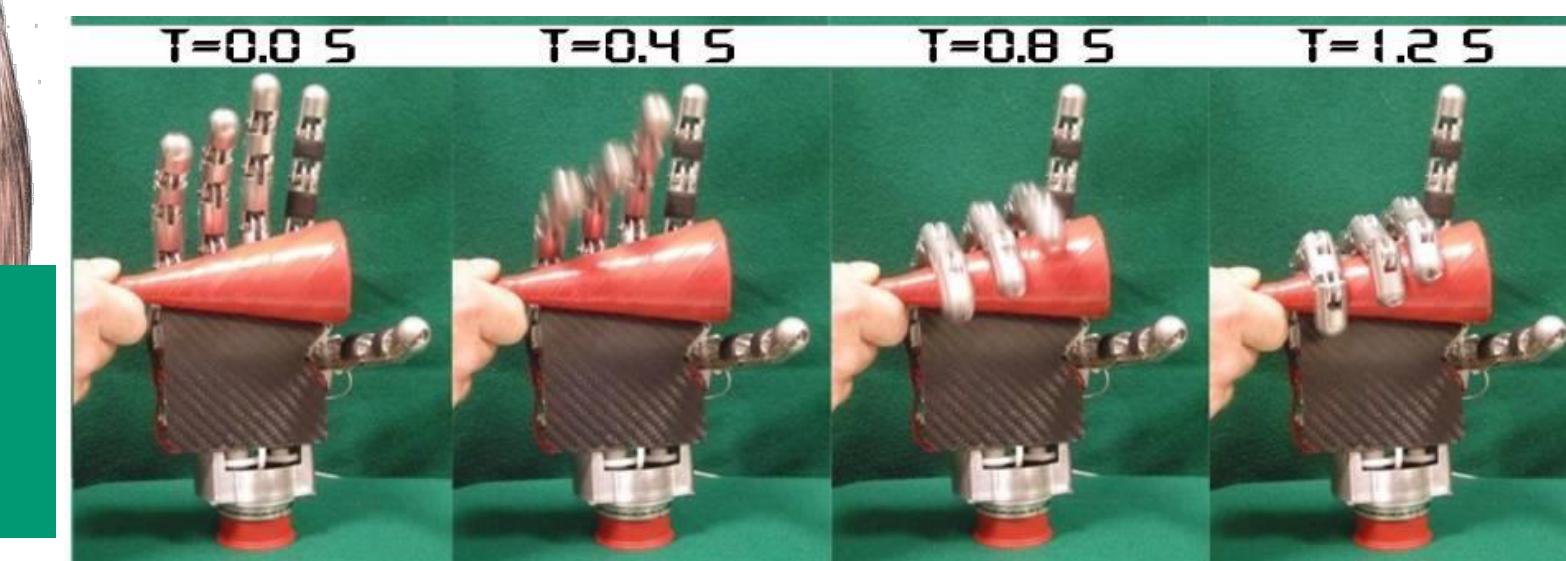
... + 18



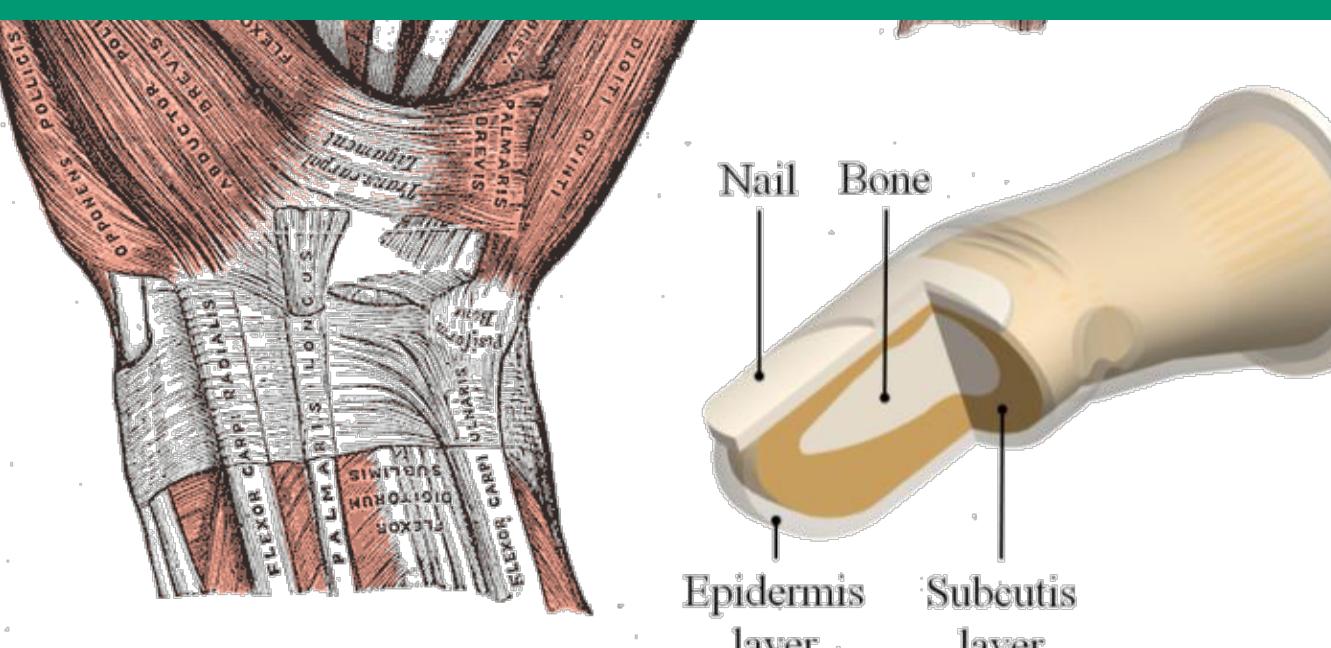
Adaptation also improves **grasp stability** as it increases the **contact areas** while grasping

Possible solutions (to simplify the problem):

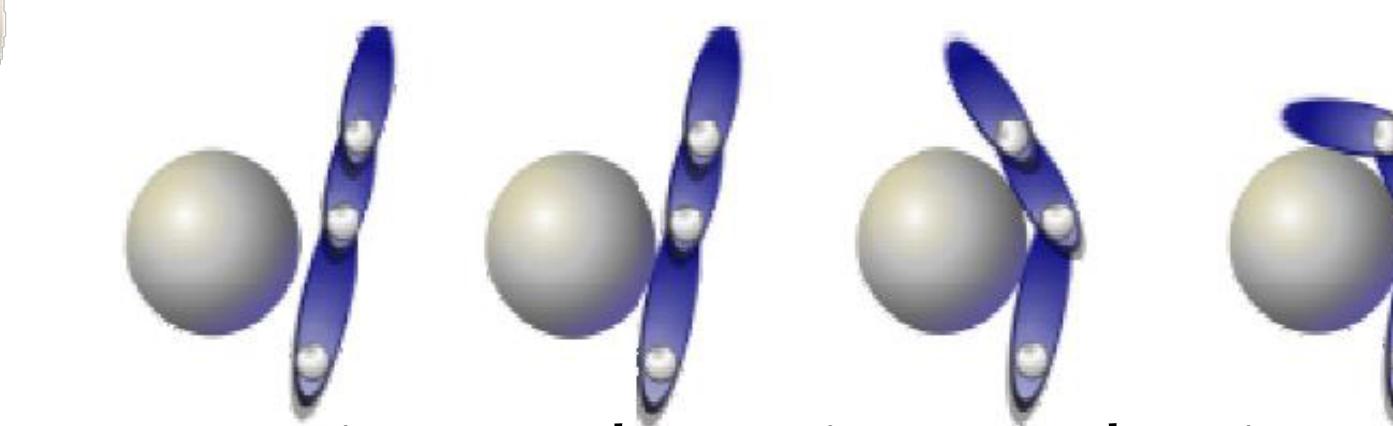
- Cut DoFs; Rigidly couple DoFs;
- Implement adaptable mechanisms.



Hand adaptation mechanisms



Phalanx adaptation
mechanisms



Finger adaptation mechanisms

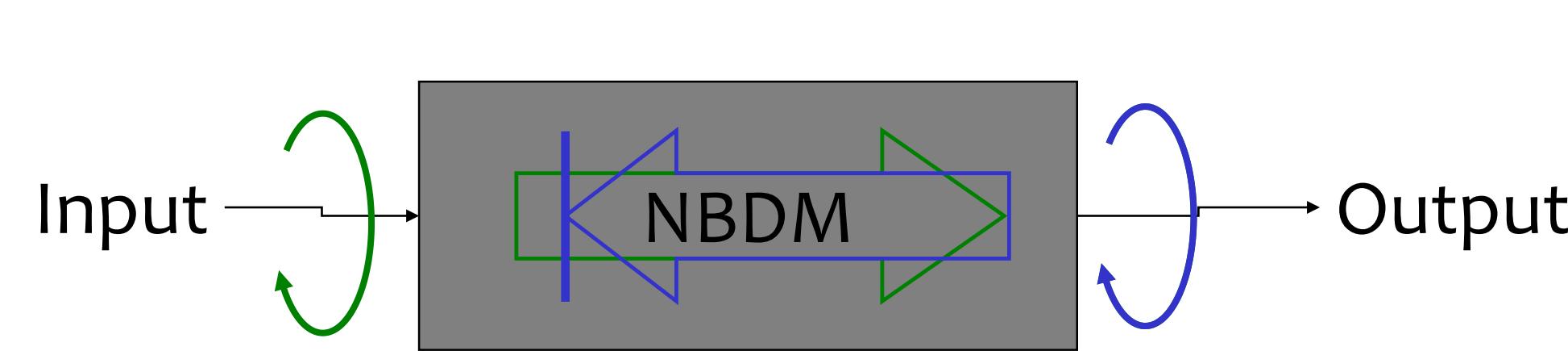
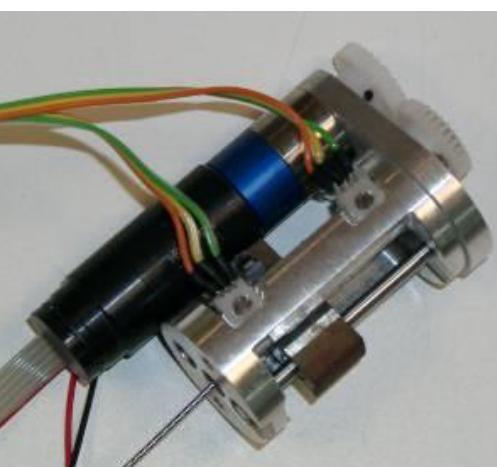


Underactuated mechanisms

Mechatronic

Design issues: non back drivability

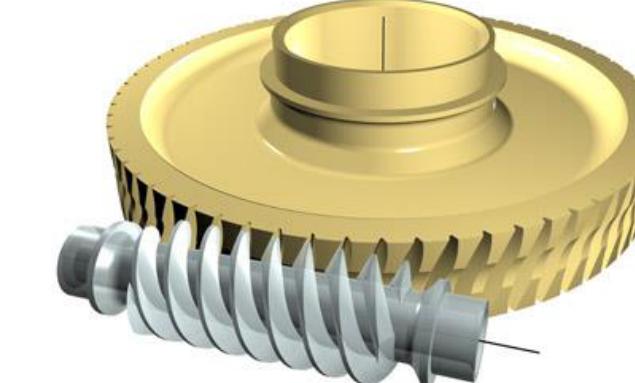
Mechanisms wherein motions generated by the input (motor) drive are **transmitted** to the output (i.e. fingers) and wherein motions originated from the output are **blocked**



In a prosthesis it allows to maintain the grasp once the power supply is switched off
Non back drivable transmission = Power saving!= key in prosthetics!



Lead Screw



Worm Gear

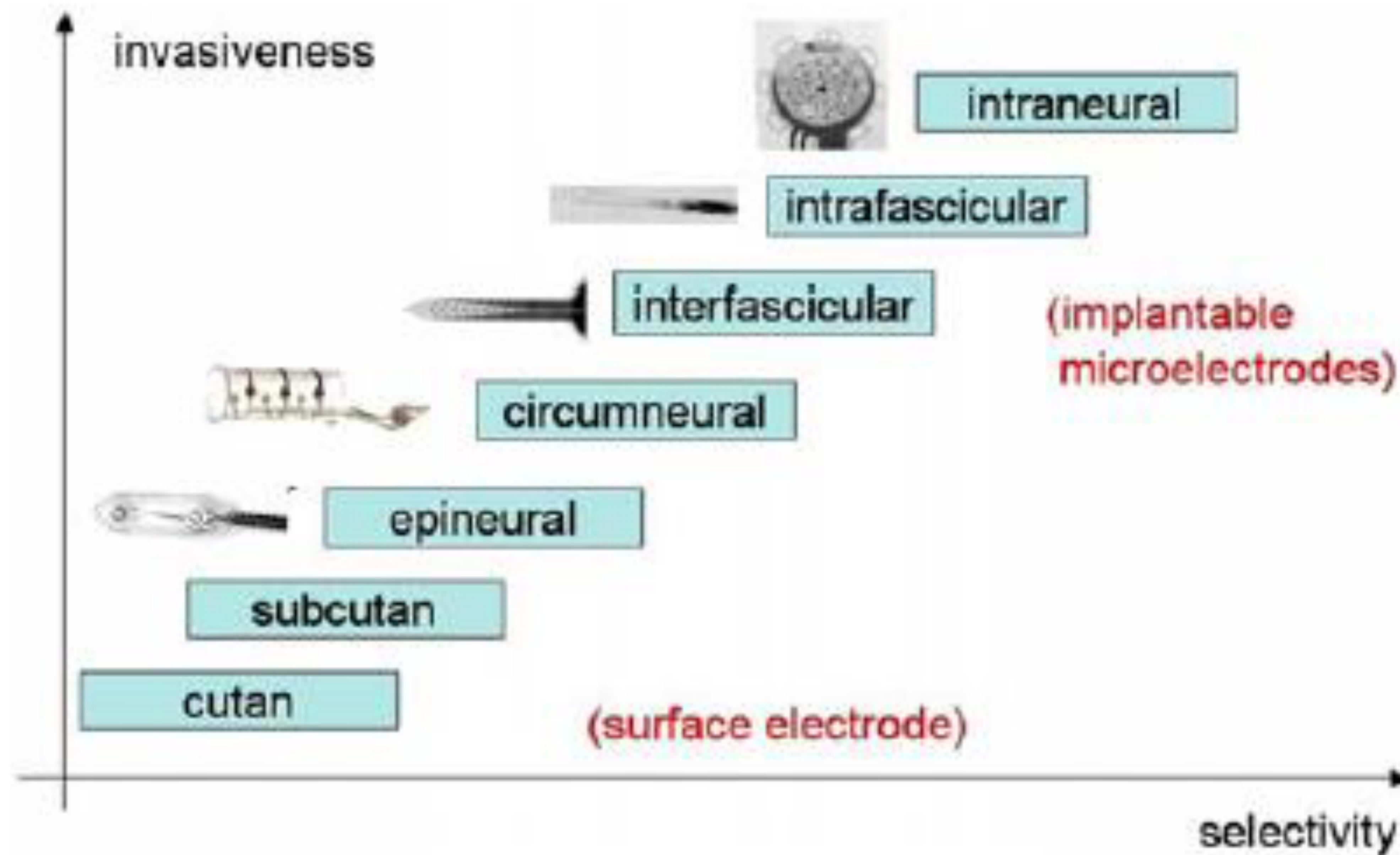


Gear heads with high reduction rate



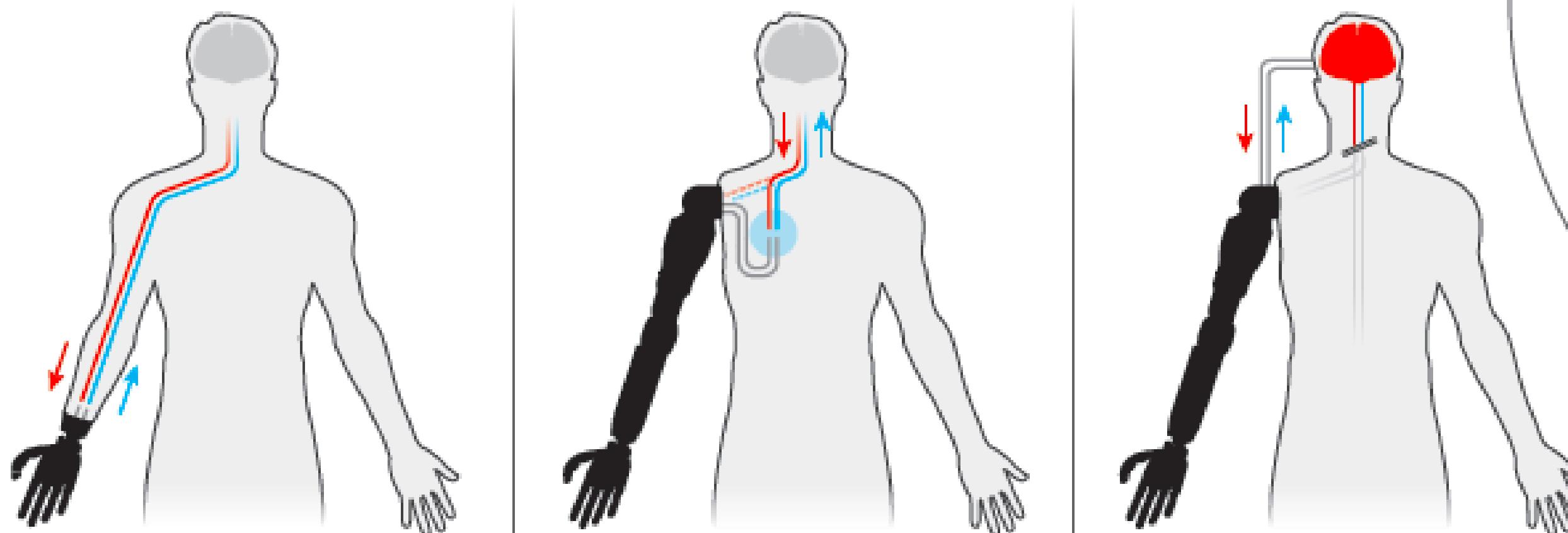
Brakes/ clutches





Sensory feedback

Real-time, and natural feedback from the hand prosthesis to the user is essential in order to enhance the control and functional impact of prosthetic hands in daily activities, prompting their full acceptance by the users



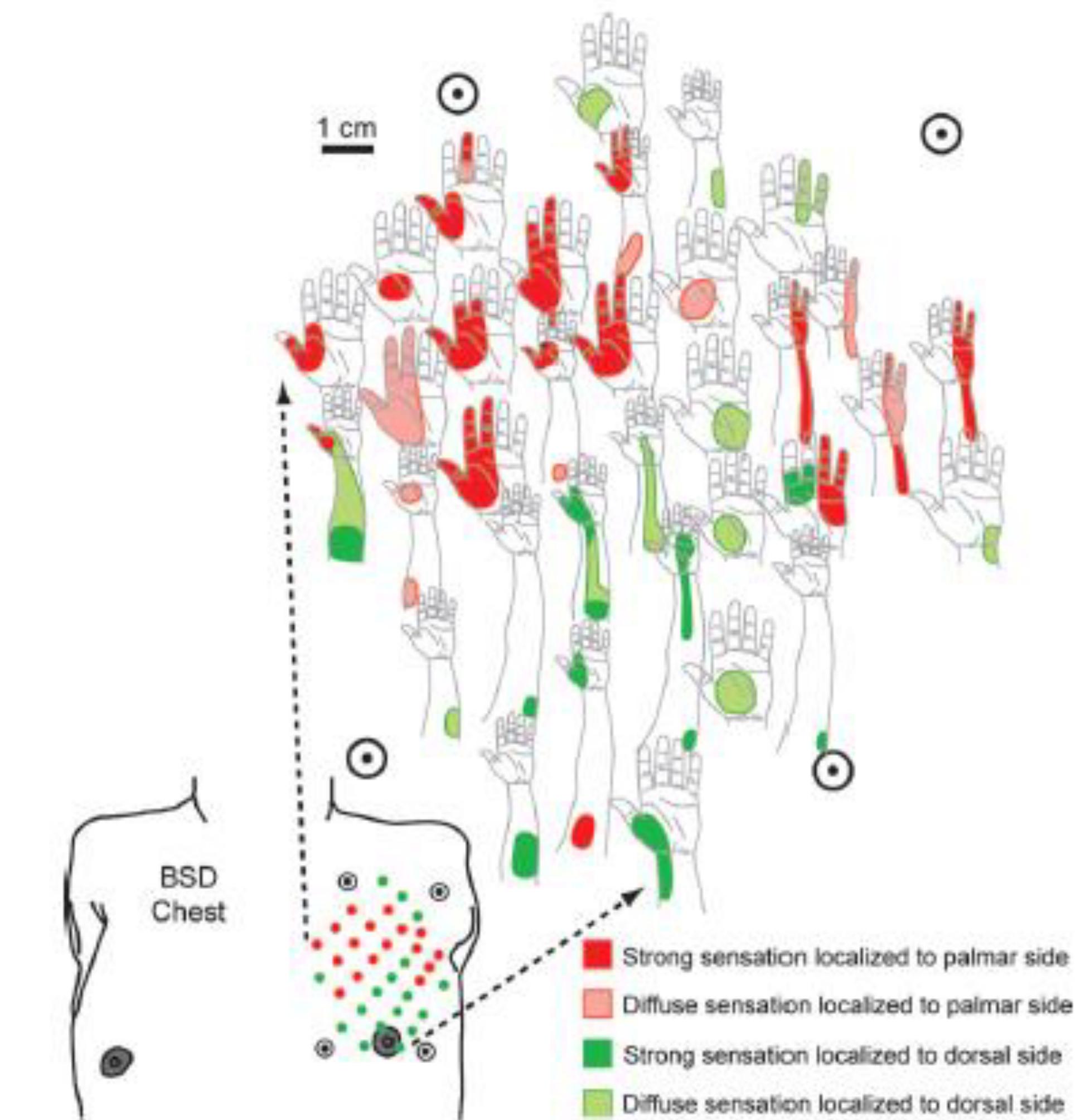
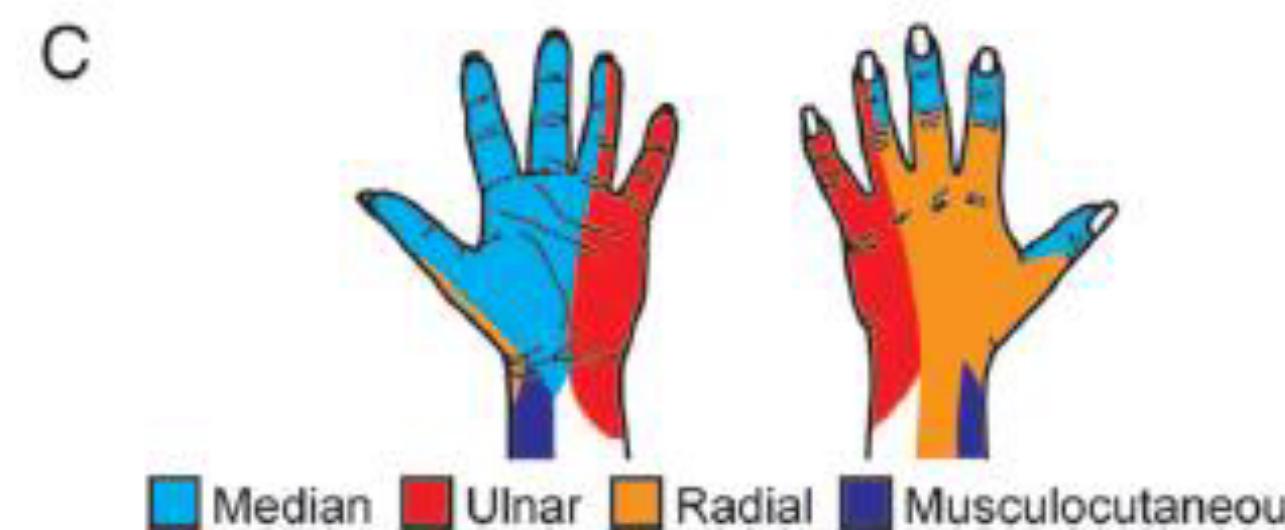
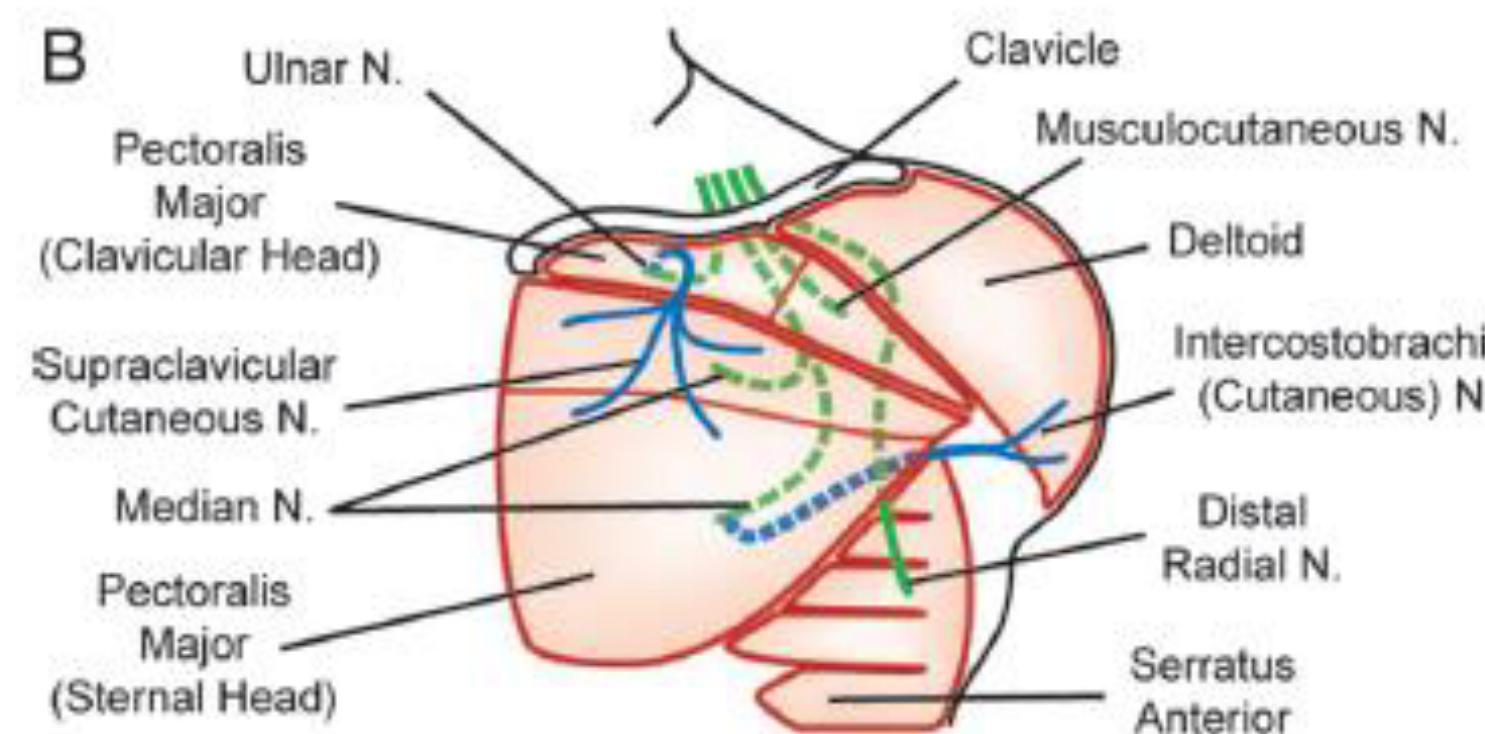
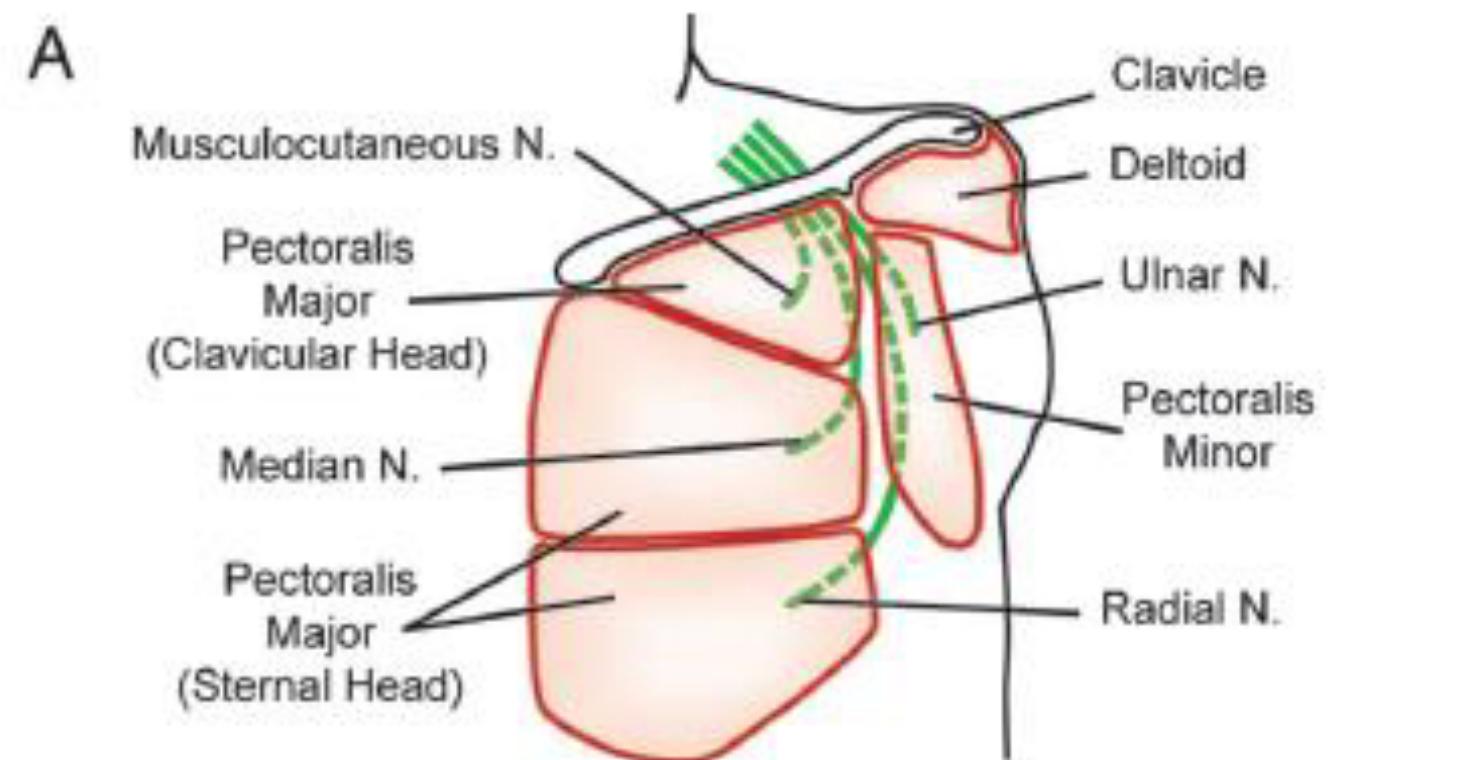
Use the remaining nerves
Electrical leads from the prosthetic's sensors stimulate nerves in the person's stump that once served the real limb.

Move the nerves
Re-routed nerves grow new endings into muscle and skin, where external devices translate signals going to and from the prosthesis.

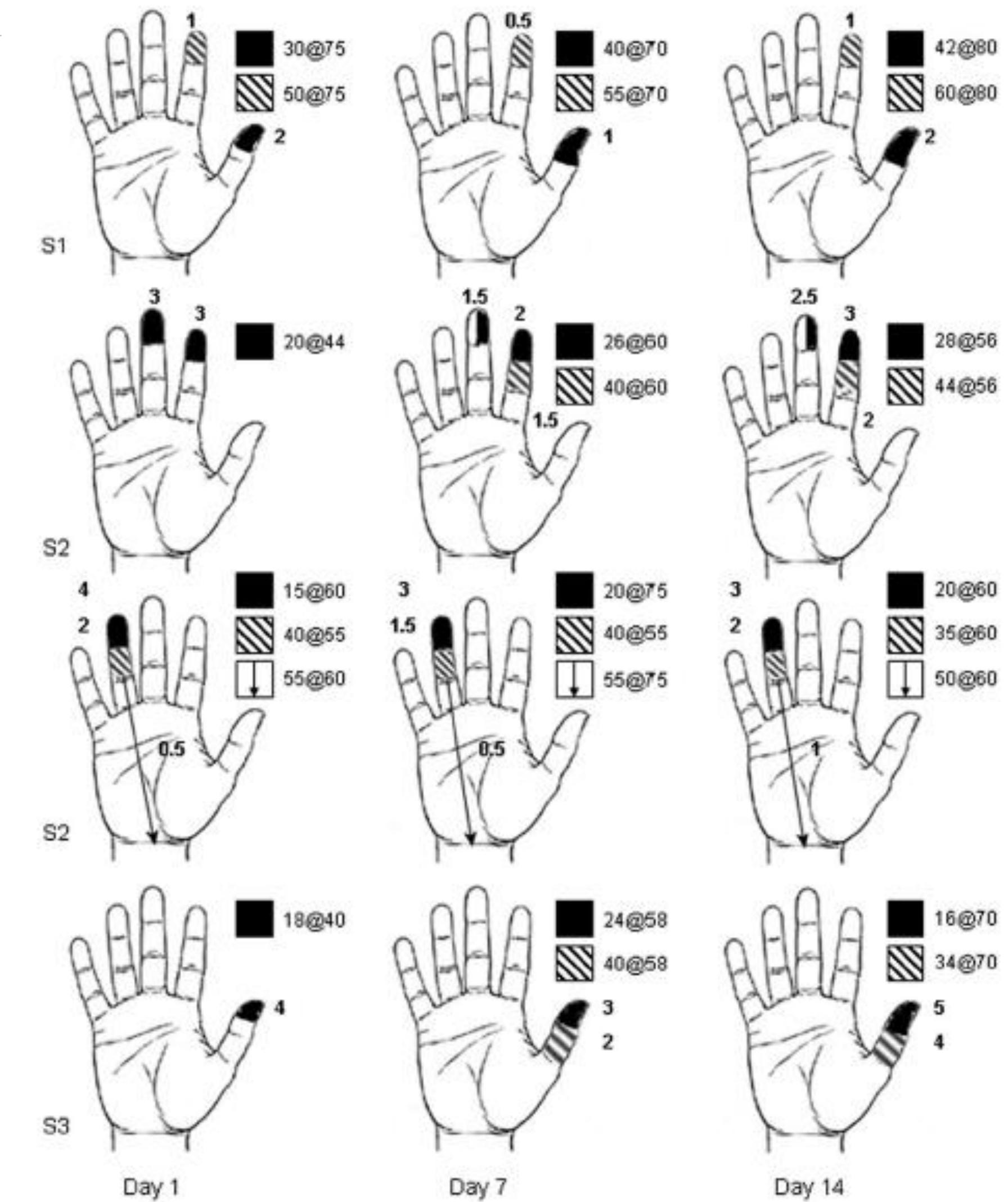
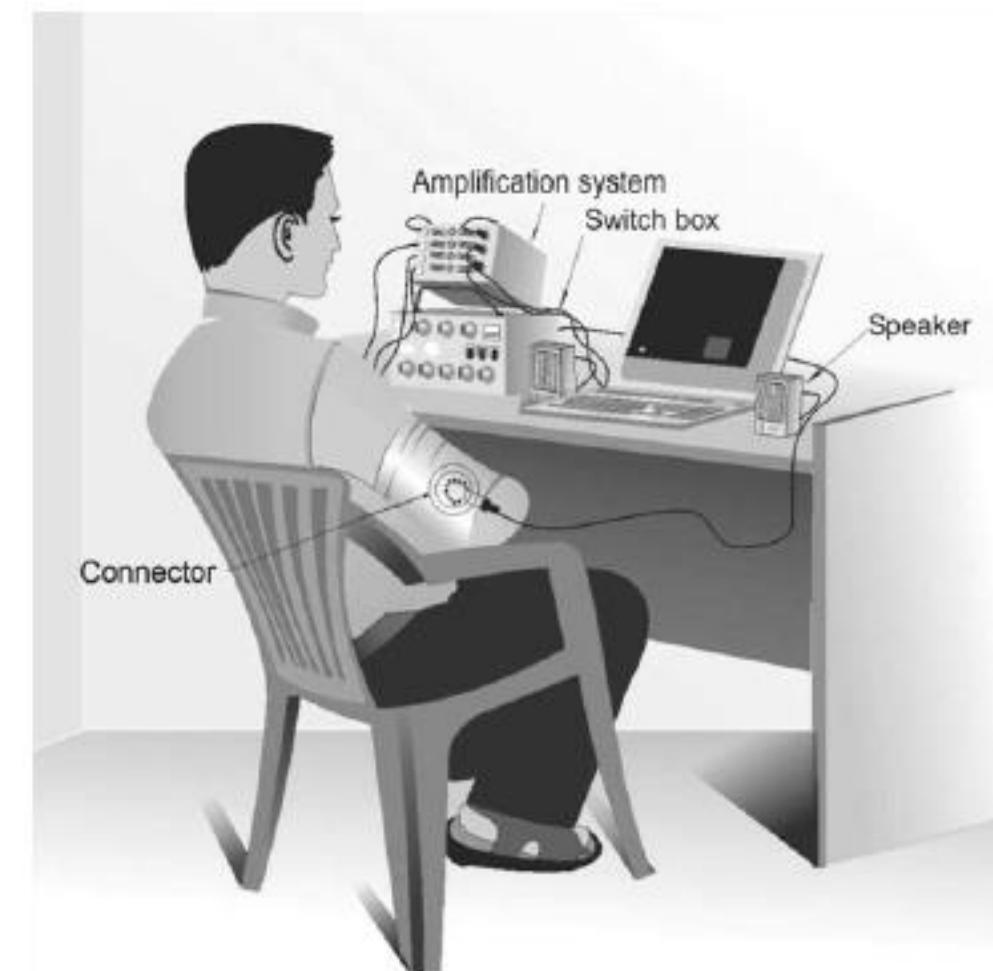
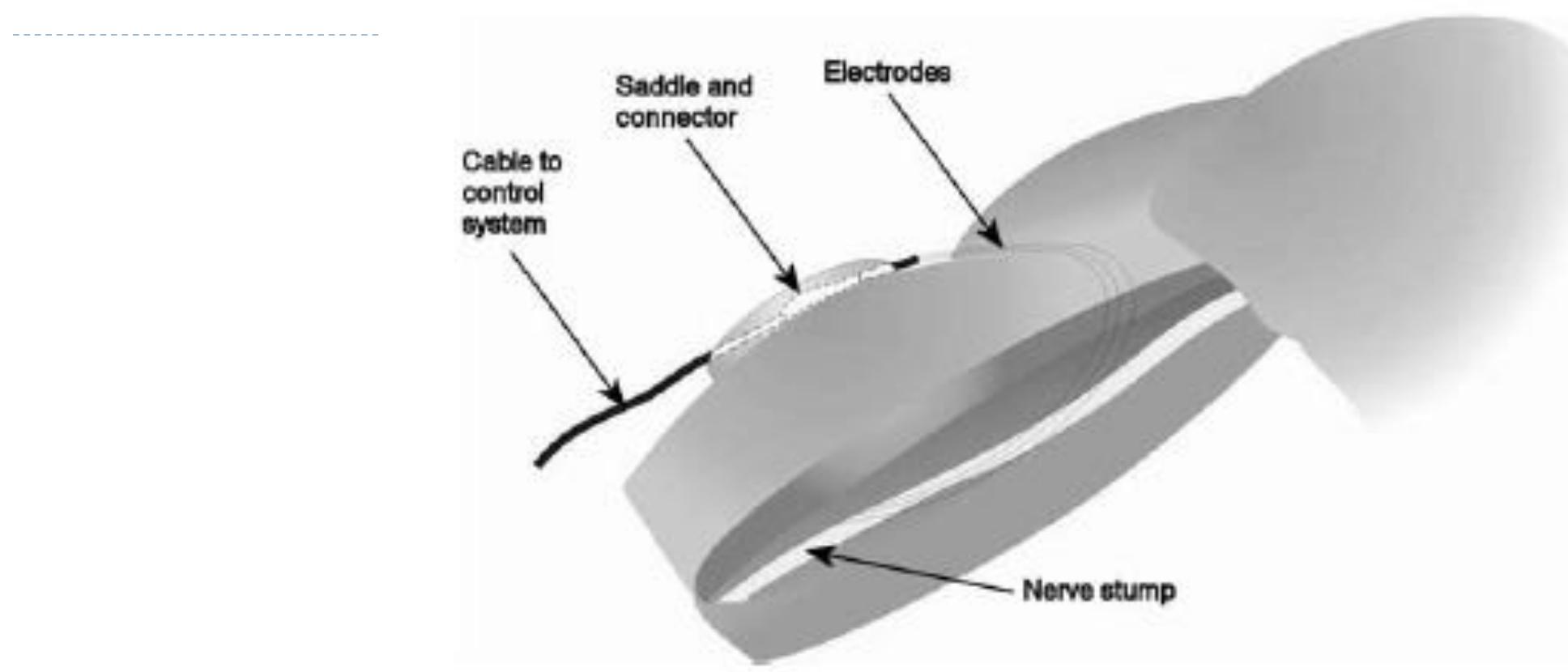
Stimulate the brain
Sensory signals are routed around a severed spinal cord and into the brain, where they produce sensations by direct stimulation of the cortex.

Kwok, Nature, 2013

Targeted Muscle Reinnervation

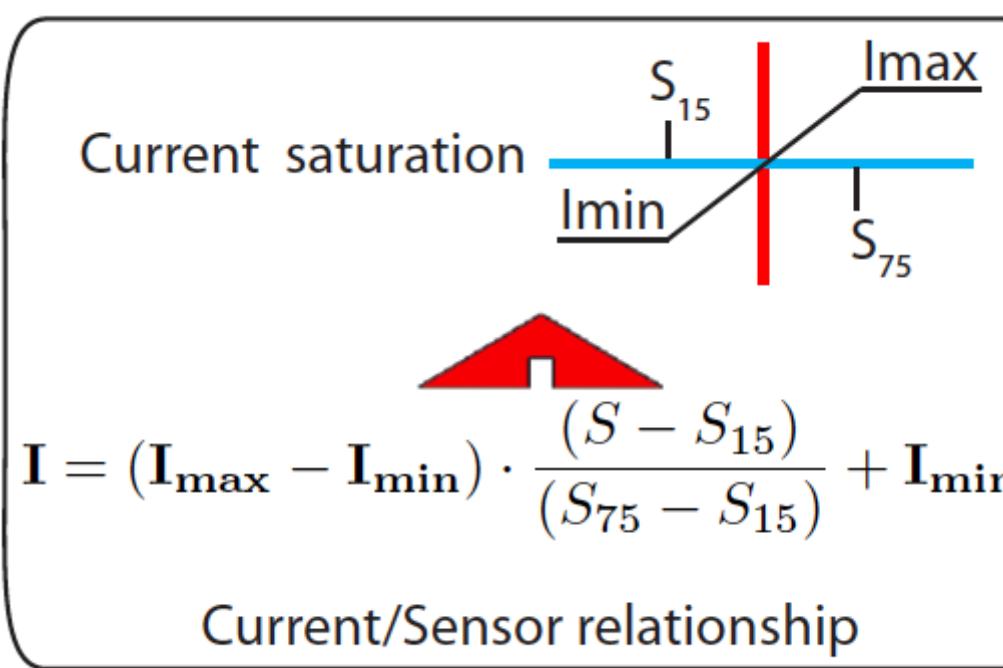
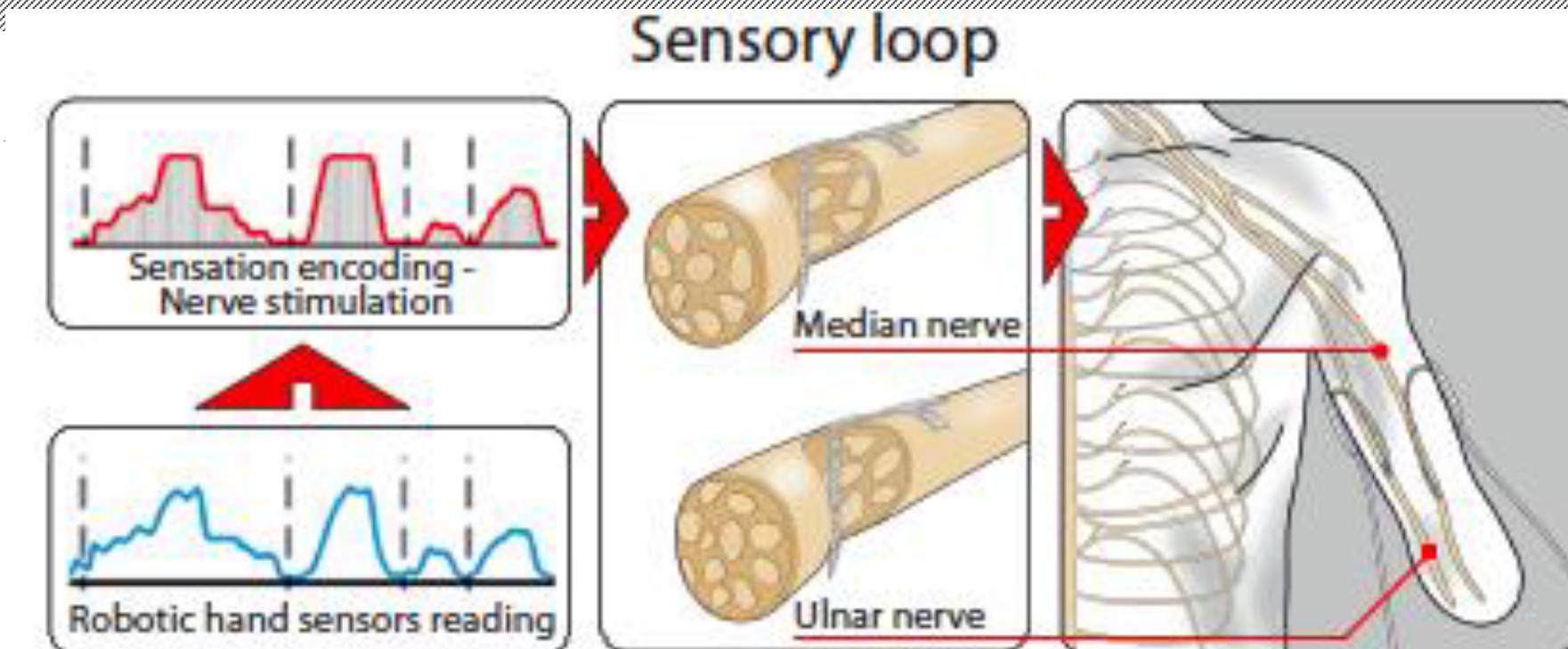


First intraneuronal experiment

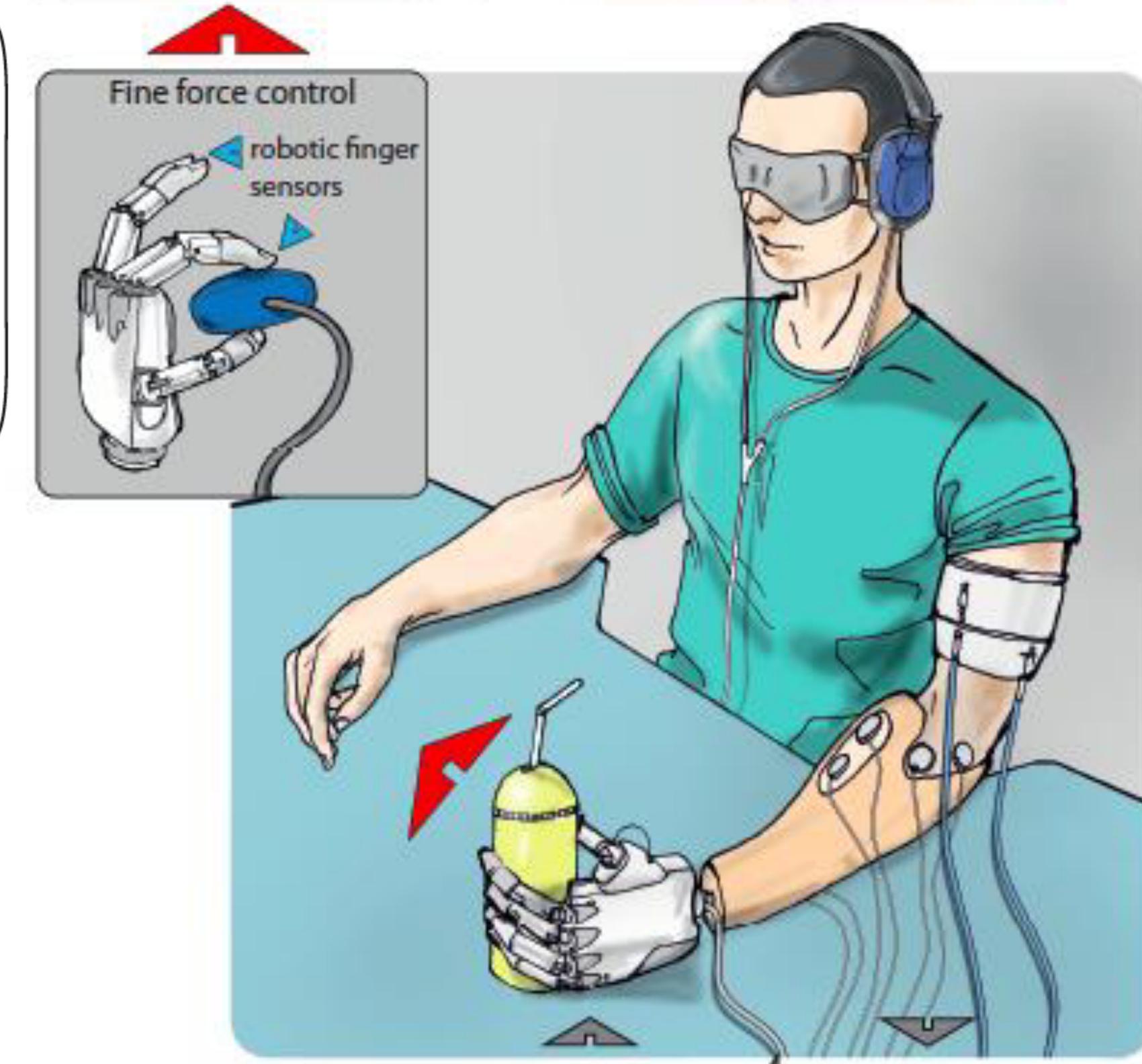


Closed-loop control based on sensory feedback

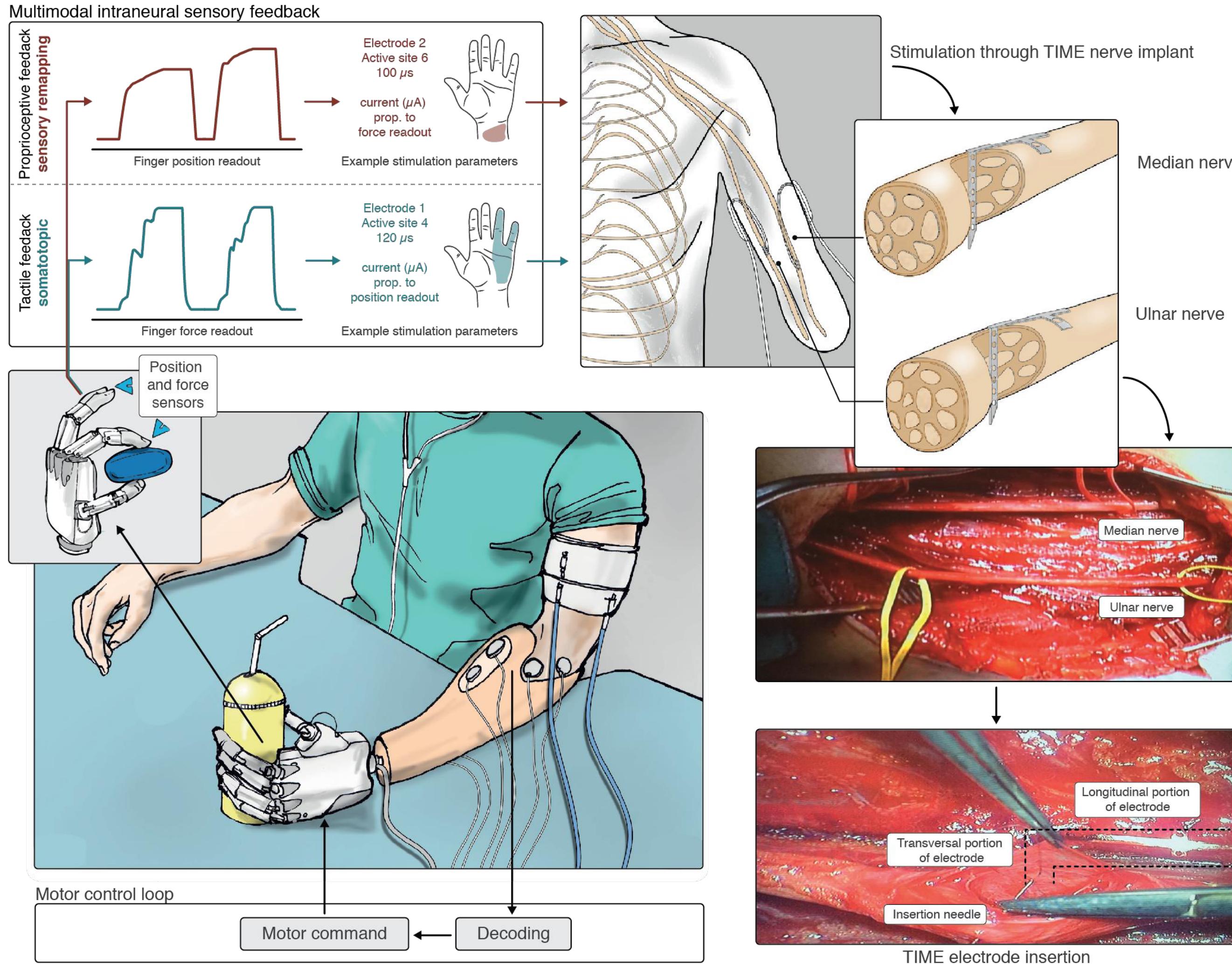
- Test the possibility for the subject to use the sensory information during closed-loop control and manipulation experiments



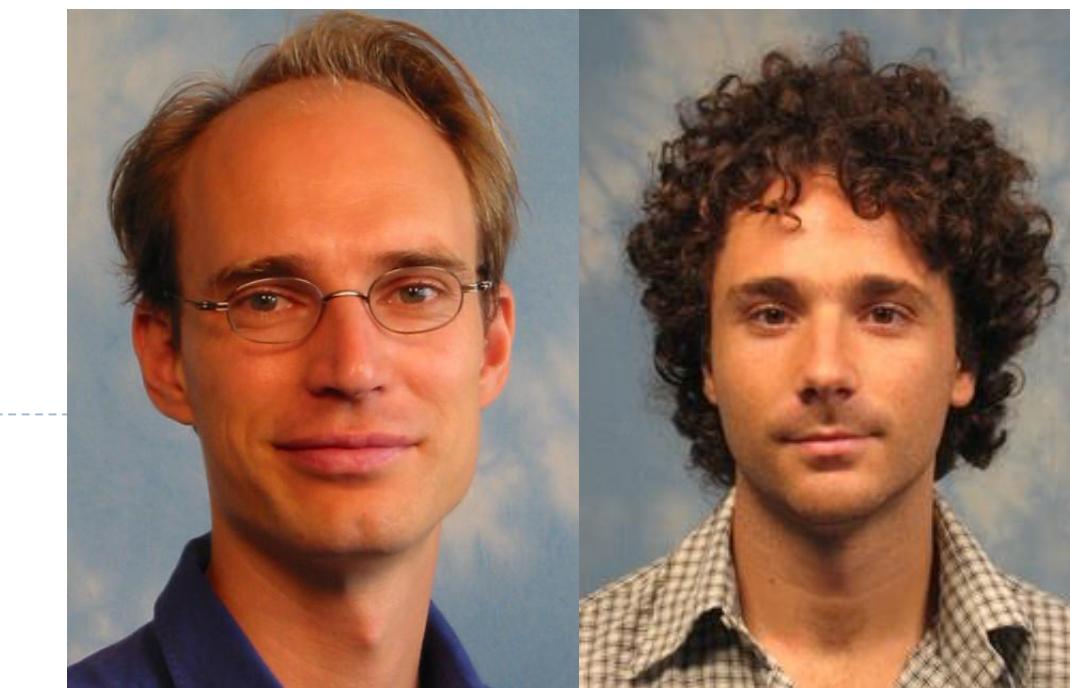
Azzurra dexterous hand
(Prensilia srl)



Restoration of proprioception and tactile feedback

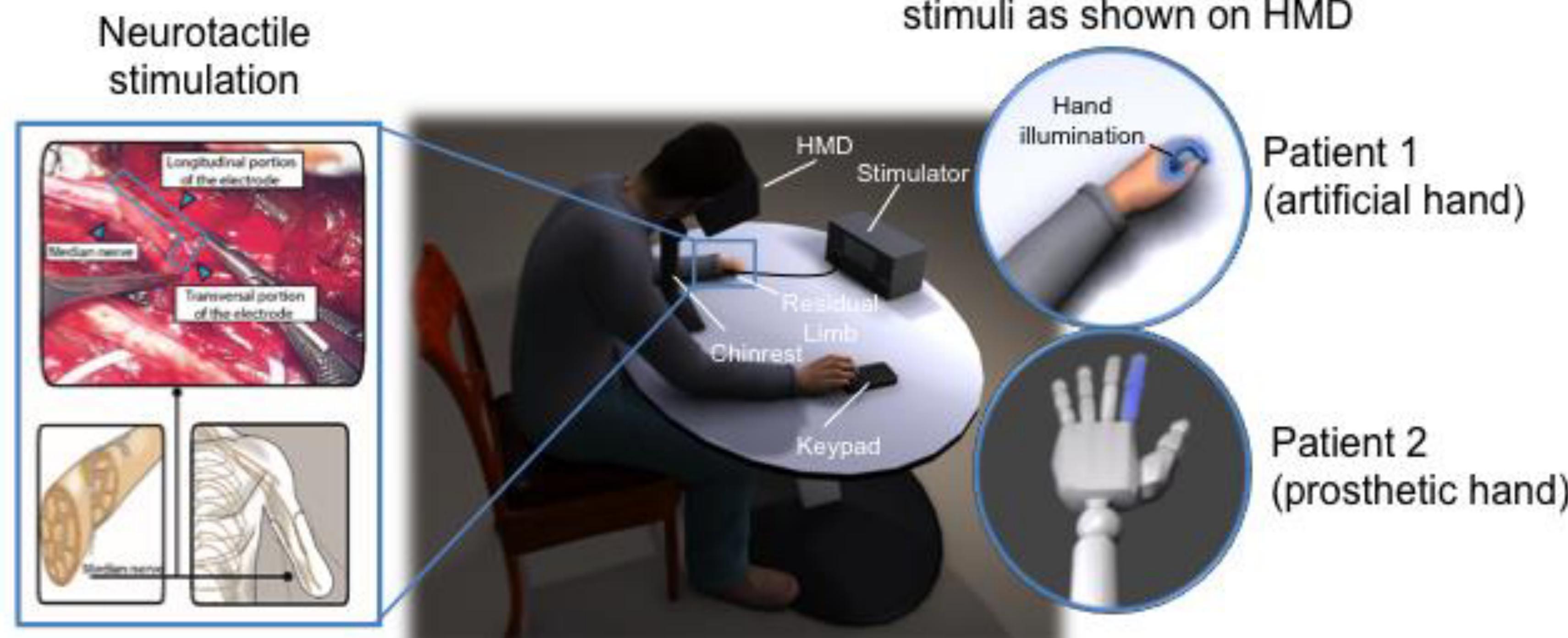


Embodiment

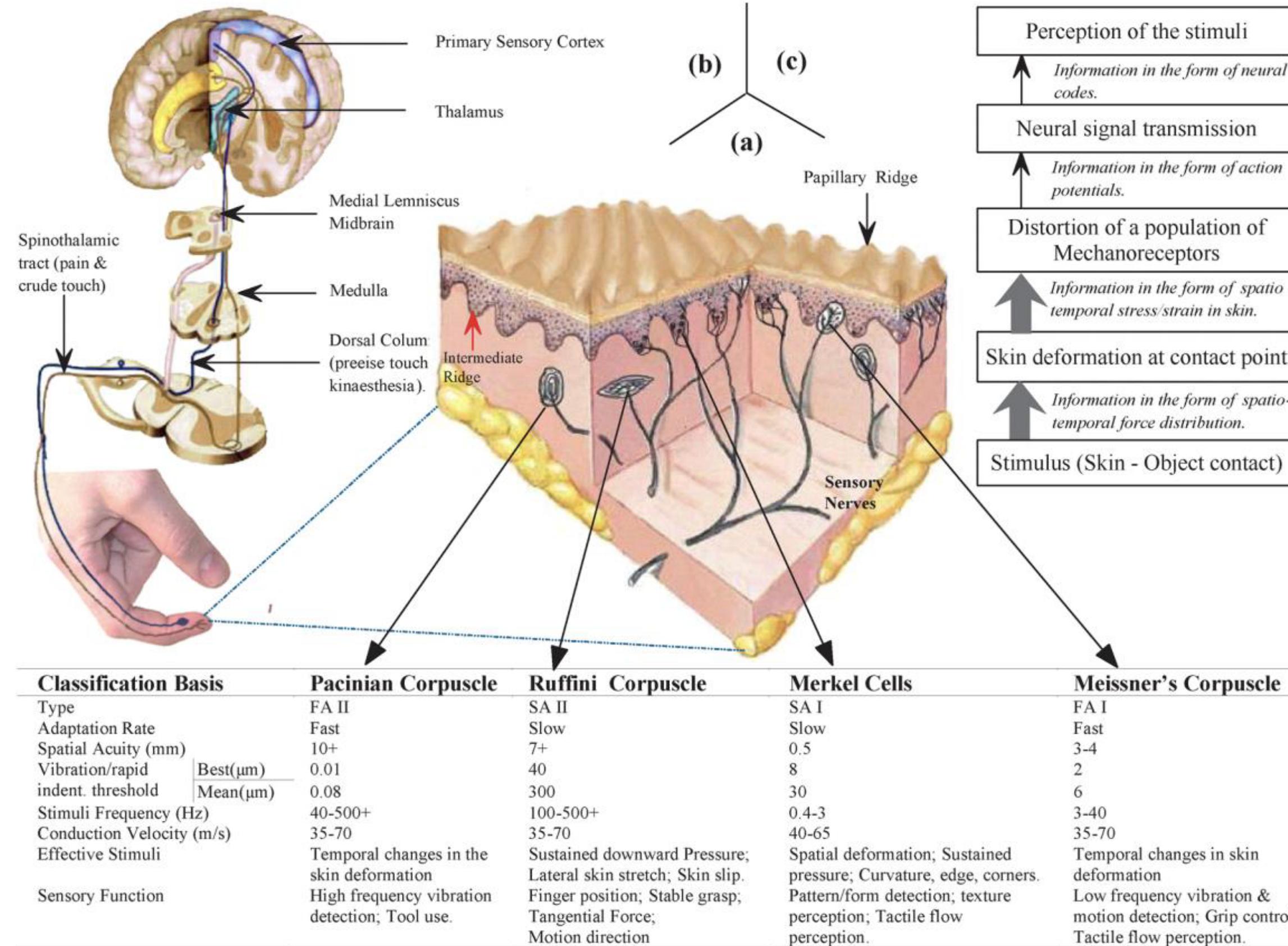


O. Blanke

G. Rognini



Human touch system

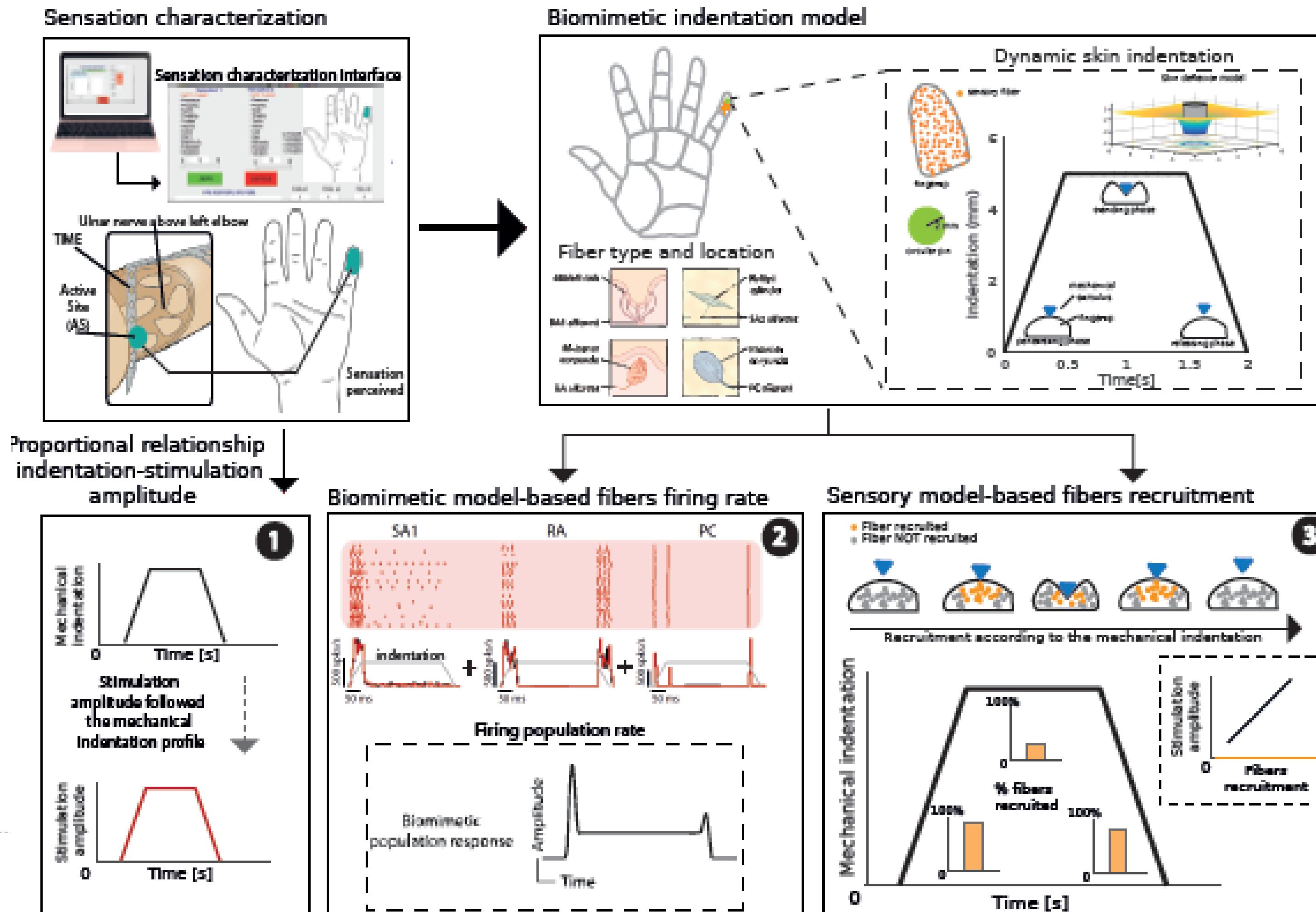


- Natural sensors fibers convey detailed information about contact events and provide us with an exquisite sensitivity to the form and surface properties of grasped objects
- During object manipulation and tactile exploration, the glabrous skin of the hand undergoes complex spatiotemporal mechanical deformations, which in turn, drive very precise spiking responses in individual afferents
- Coarse object features, such as edges and corners, are reflected in spatial patterns of activation in slowly adapting type I (SAI) and rapidly adapting (FA) fibers, which are densely packed in the fingertip
- At the same time, interactions with objects and surfaces elicit high-frequency, low-amplitude surface waves that propagate across the skin of the finger and palm and excite vibration-sensitive Pacinian (PC) afferents all over the hand

Biomimetic encoding strategy

Step 1: Biomimetic model-based approach and parameters generation

We identified electrode active site which elicits sensations in the locations corresponding to the fingertip. Then, we simulated a mechanical skin indentation using the biomimetic model. The model outcomes were the firing population activity generated by the combination of all the fibers (SA, RA, PC) response and the number of sensory fibers recruited during the skin indentation. We also generated the stimulation amplitudes following a proportional relationship with the mechanical stimulus as used in (16).



Biomimetic encoding strategy

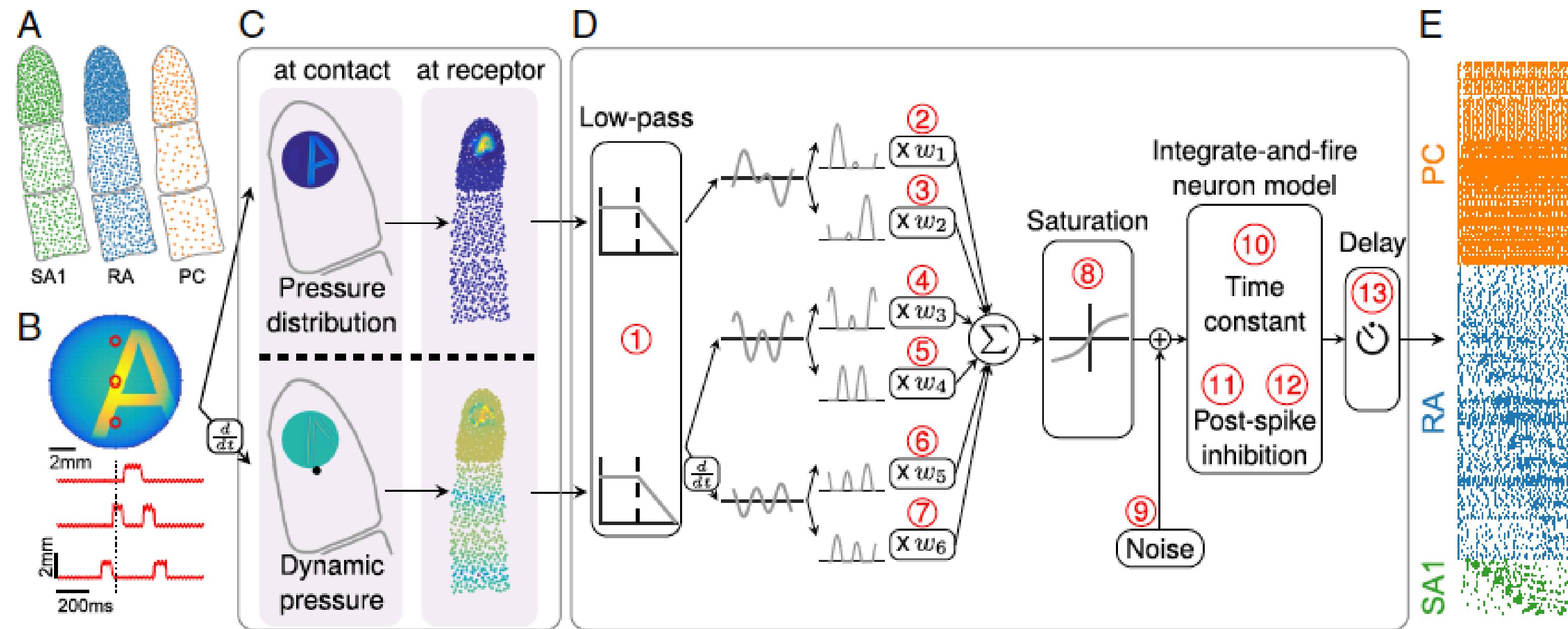


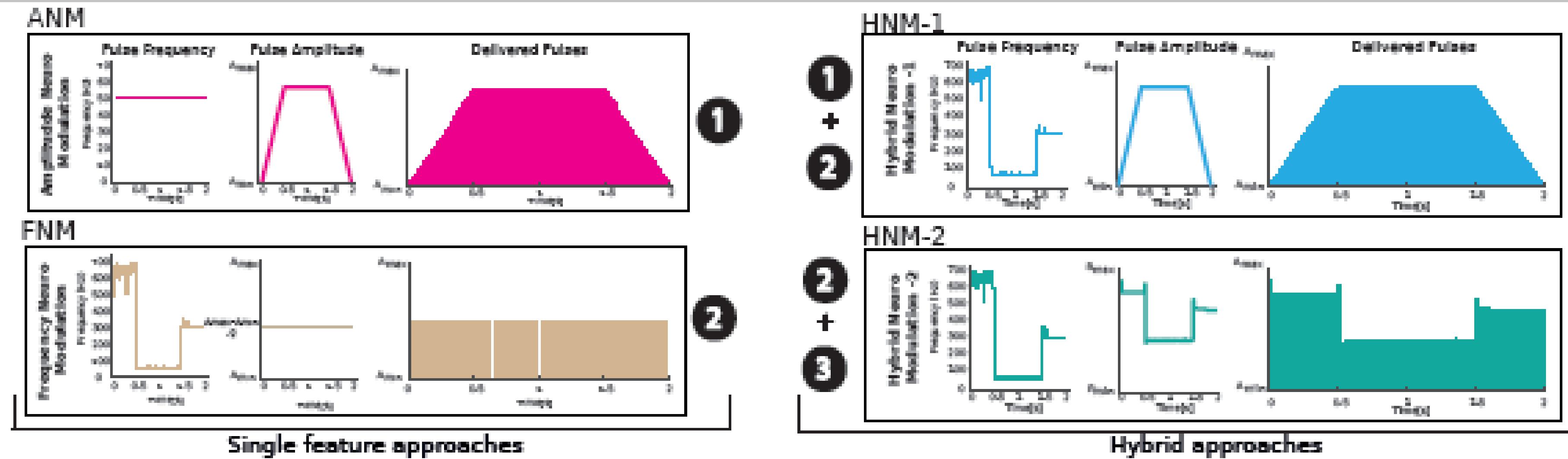
Fig. 1. Overview of the model. (A) Receptors are distributed across the skin given the known innervation densities of SA1, RA, and PC afferents. (B) The stimulus—in this case, a vibrating embossed letter A scanned across the skin—is defined as the time-varying depth at which each small patch of skin (here dubbed a pin) is indented (with a spatial resolution of 0.1 mm). The traces in *Lower* show the time-varying depth at the three locations on the skin indicated by the red dots in *Upper*. (C) The mechanics model relies on two parts: (*Upper*) modeling the distribution of stresses using a quasistatic elastic model and (*Lower*) modeling dynamic pressure and surface wave propagation. *Left* shows the surface deformation of the skin, and *Right* shows the resulting pattern of stresses at the location of the receptors. (D) The spiking responses are determined by leaky IF models using different sets of up to 13 parameters (marked in red numbers) for individual SA1, RA, and PC afferents fit based on peripheral recordings to skin vibrations. Adapted from ref. 71. (E) The output of the model is the spike train of each afferent in the population. Raster of the response of the afferent population sampled as in A to the stimulus shown in B (only active afferents are included). Note that the SA1s (in contact) only encode the spatial aspect of the stimulus, that the PCs encode from the whole finger phase-lock with the 200-Hz vibration, and that the RAs show mixed spatial and vibration responses.



Biomimetic encoding strategy

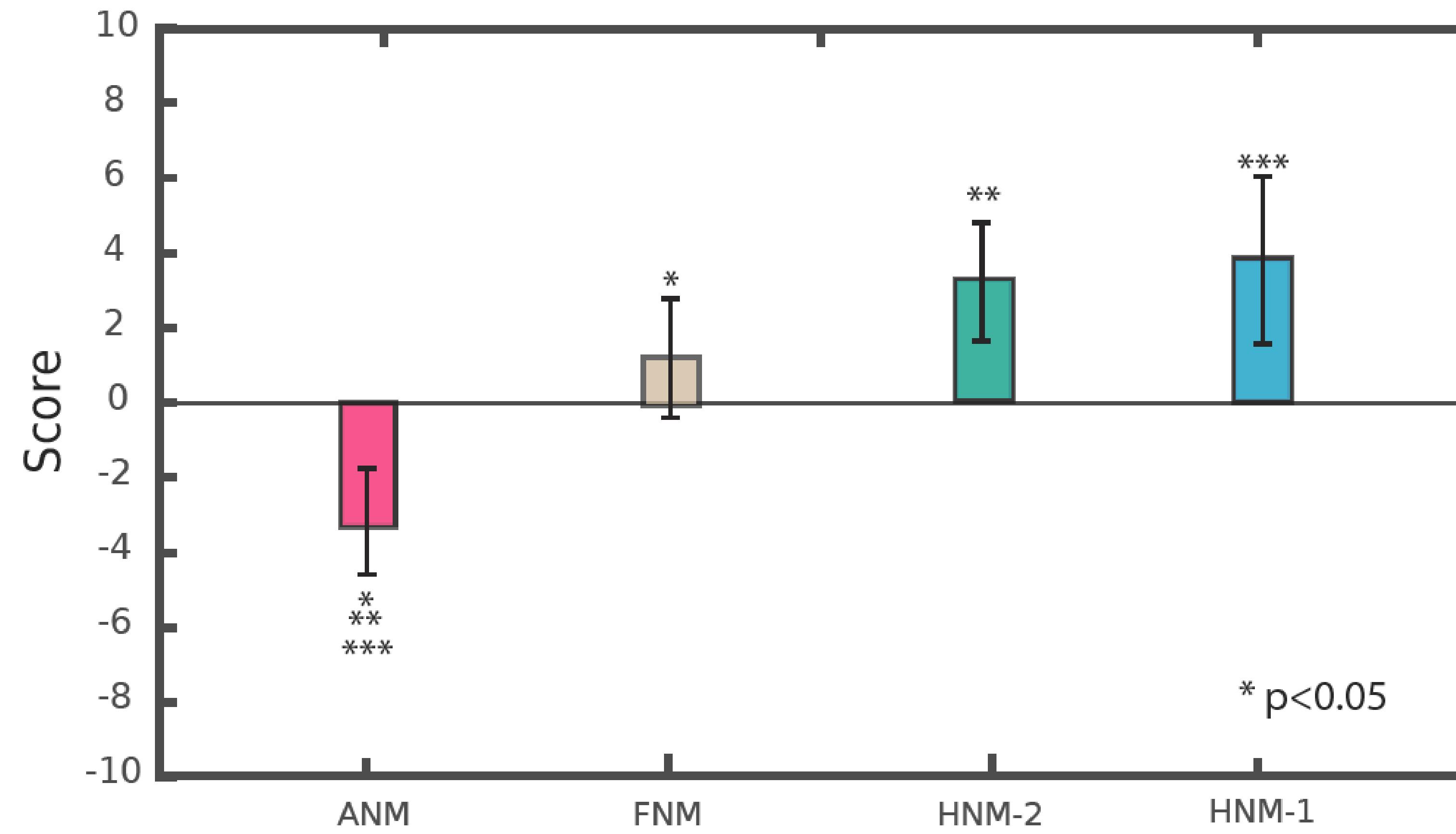
Step 2: Sensory encoding strategies

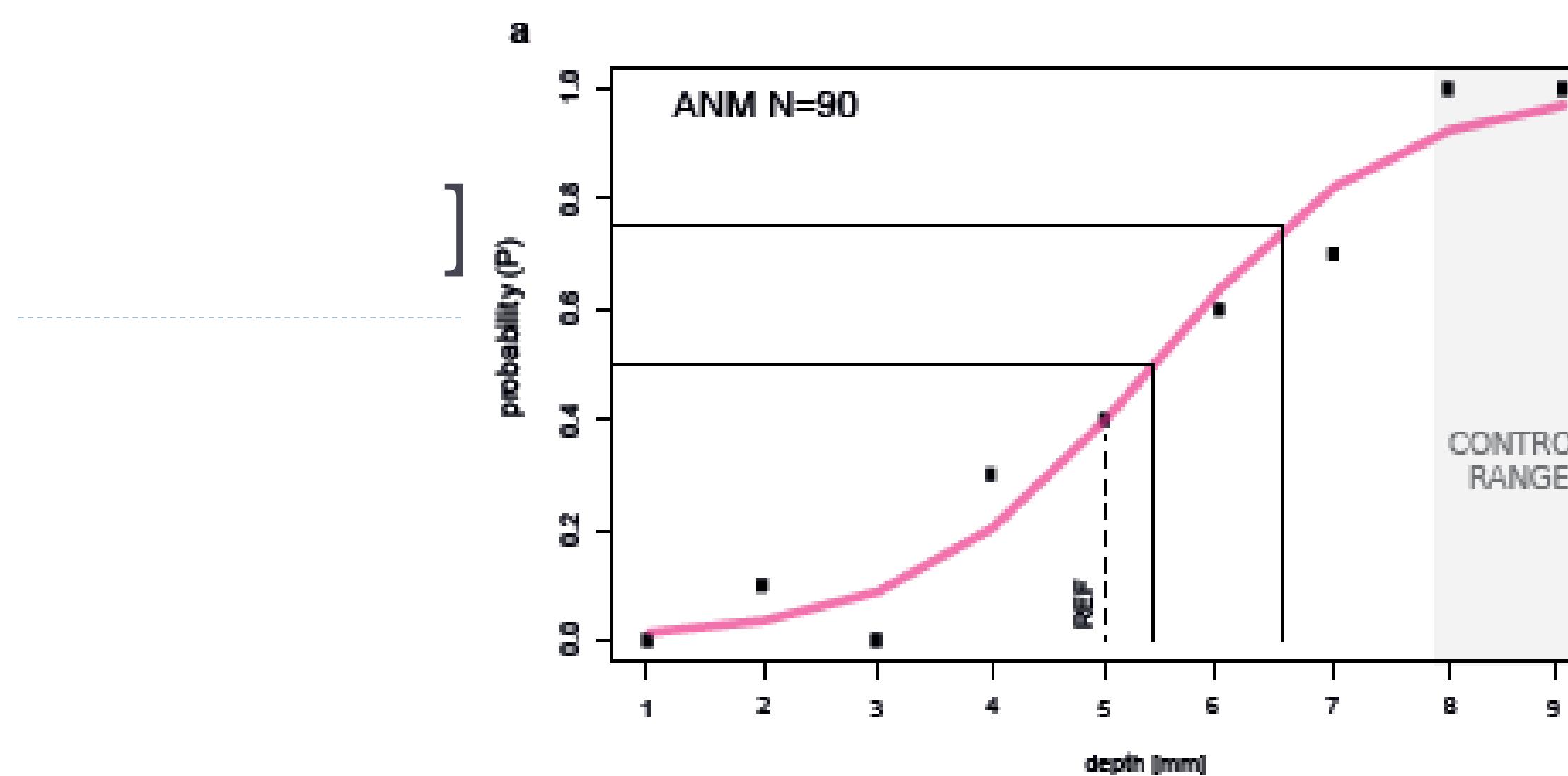
Different encoding strategies in which only one stimulation feature is modulated (Single feature) or both frequency and amplitude of the stimuli are simultaneously modulated (Hybrid). We converted the firing population rate generated by the biomimetic model in the frequency of the intraneuronal stimulation (FNM, HNM-1 and HNM-2). The stimulation amplitude was converted using the mechanical stimulus (ANM) and HNM-1) or the fibers recruitment (HNM-2). The pulse-width was always fixed to 60 μ s.



Biomimetic encoding strategy

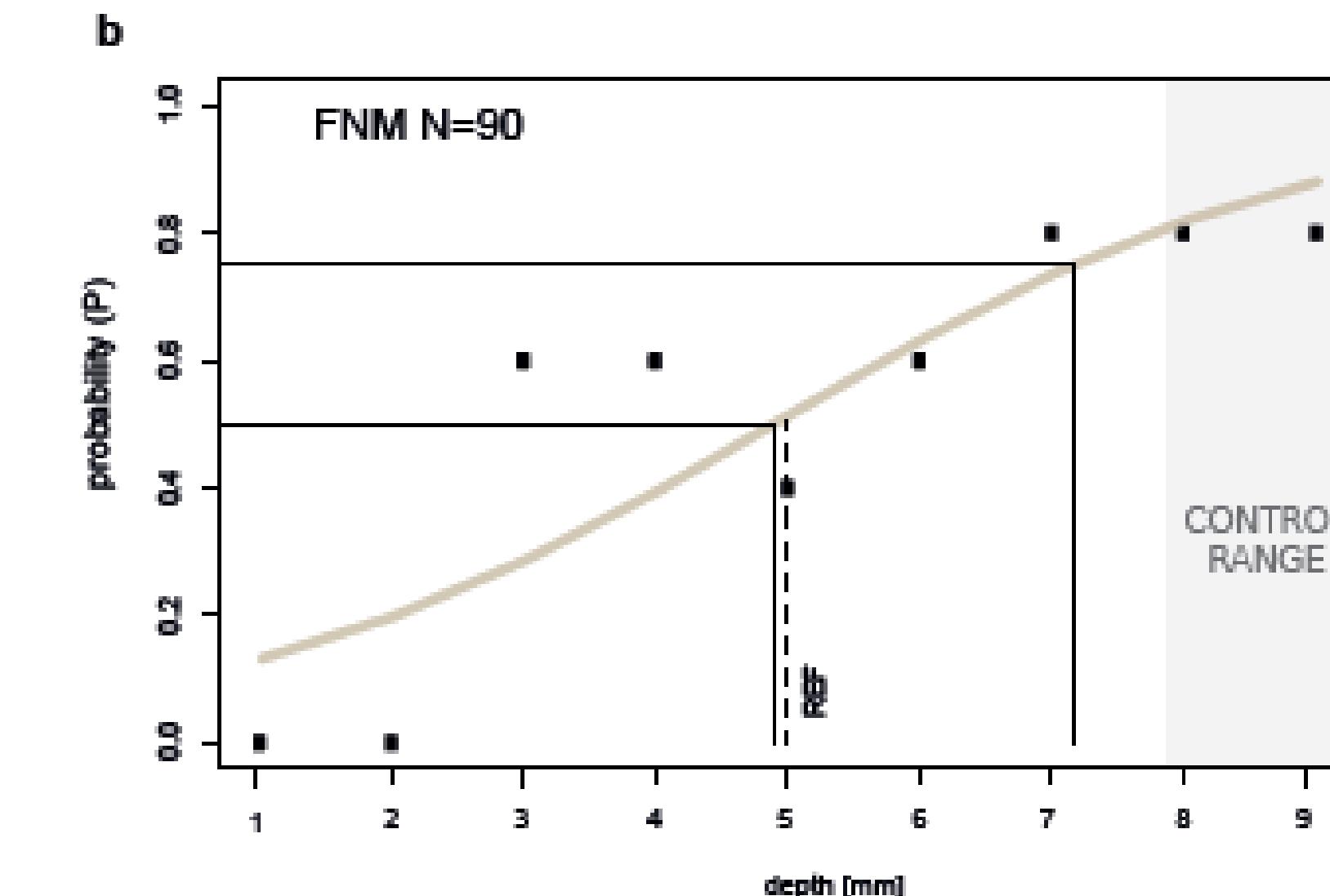
b Perceived naturalness among different encoding strategies N=16





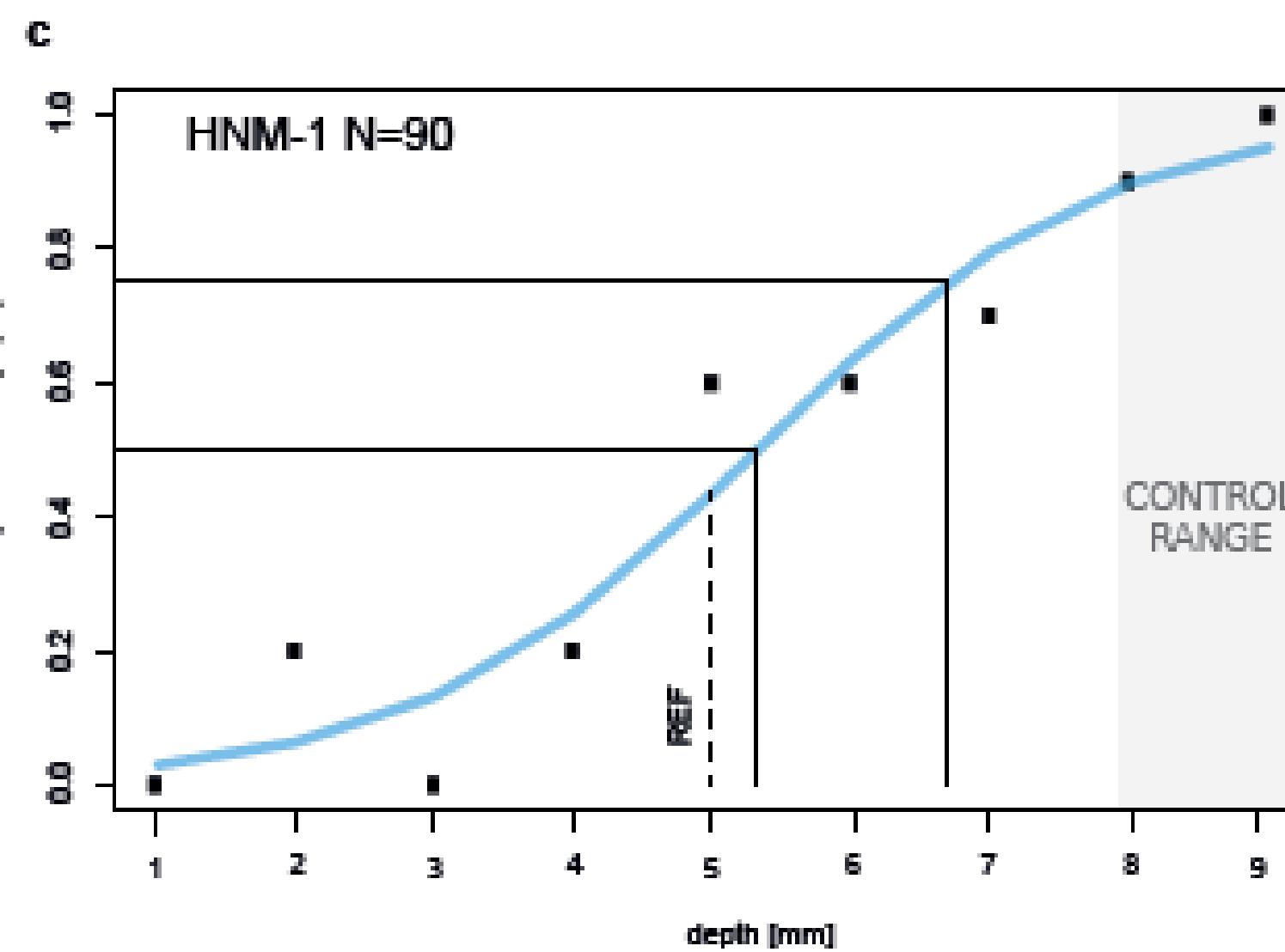
Point of Subjective Equality (PSE): 5.51 mm

Just-Noticeable Difference (JND): 1.01 mm



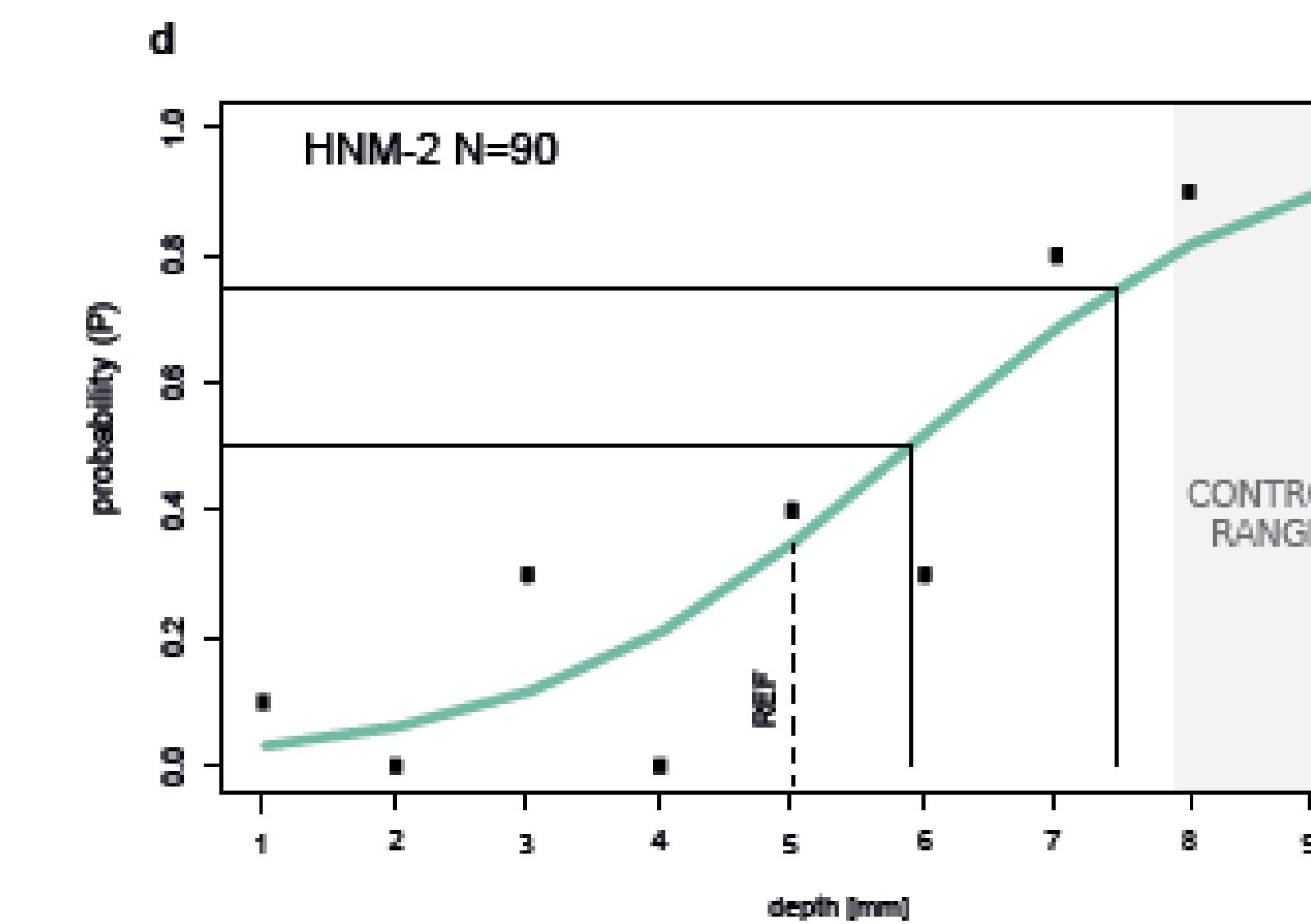
Point of Subjective Equality (PSE): 4.87 mm

Just-Noticeable Difference (JND): 2.26 mm



Point of Subjective Equality (PSE): 5.31 mm

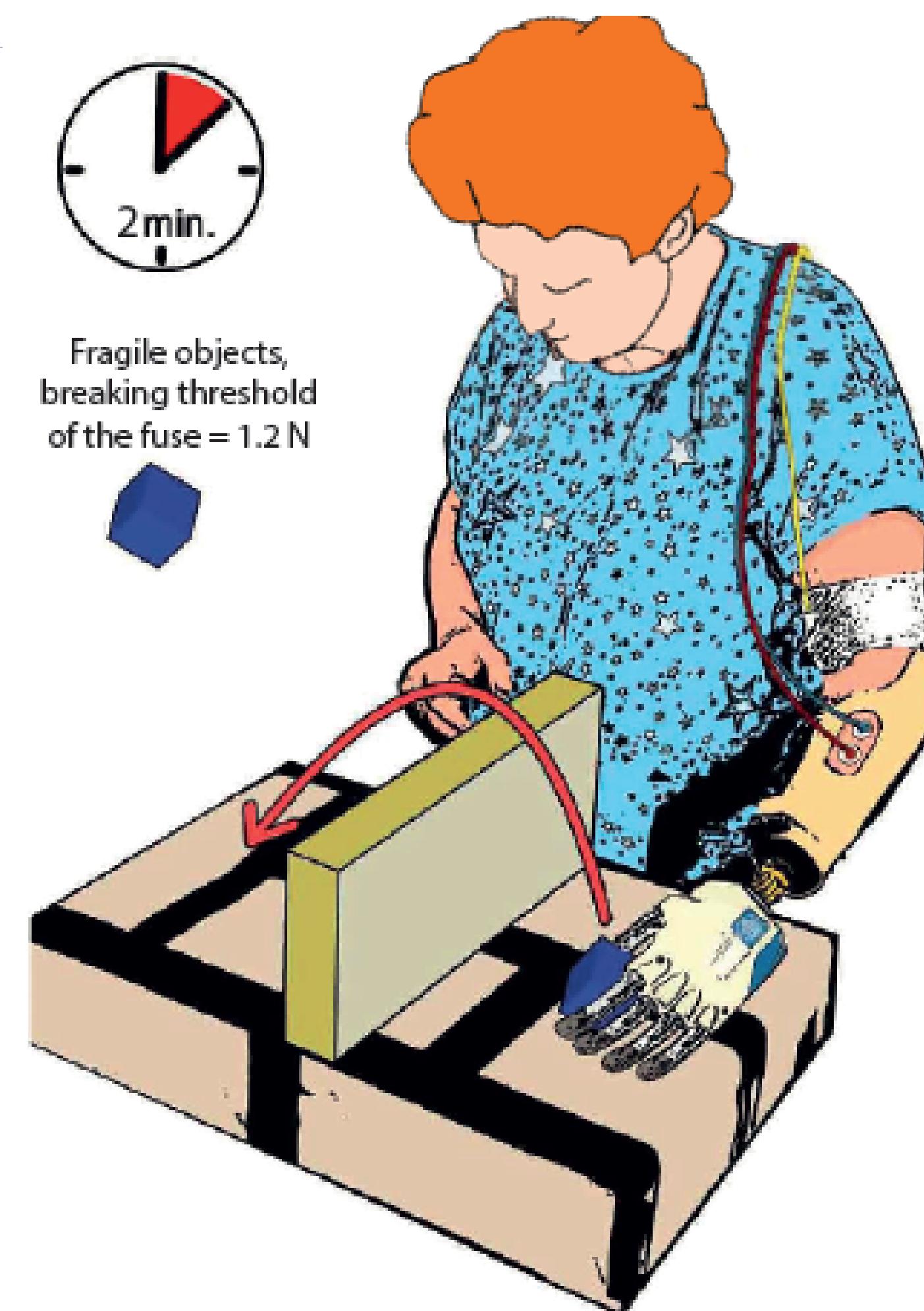
Just-Noticeable Difference (JND): 1.35 mm



Point of Subjective Equality (PSE): 5.87 mm

Just-Noticeable Difference (JND): 1.55 mm

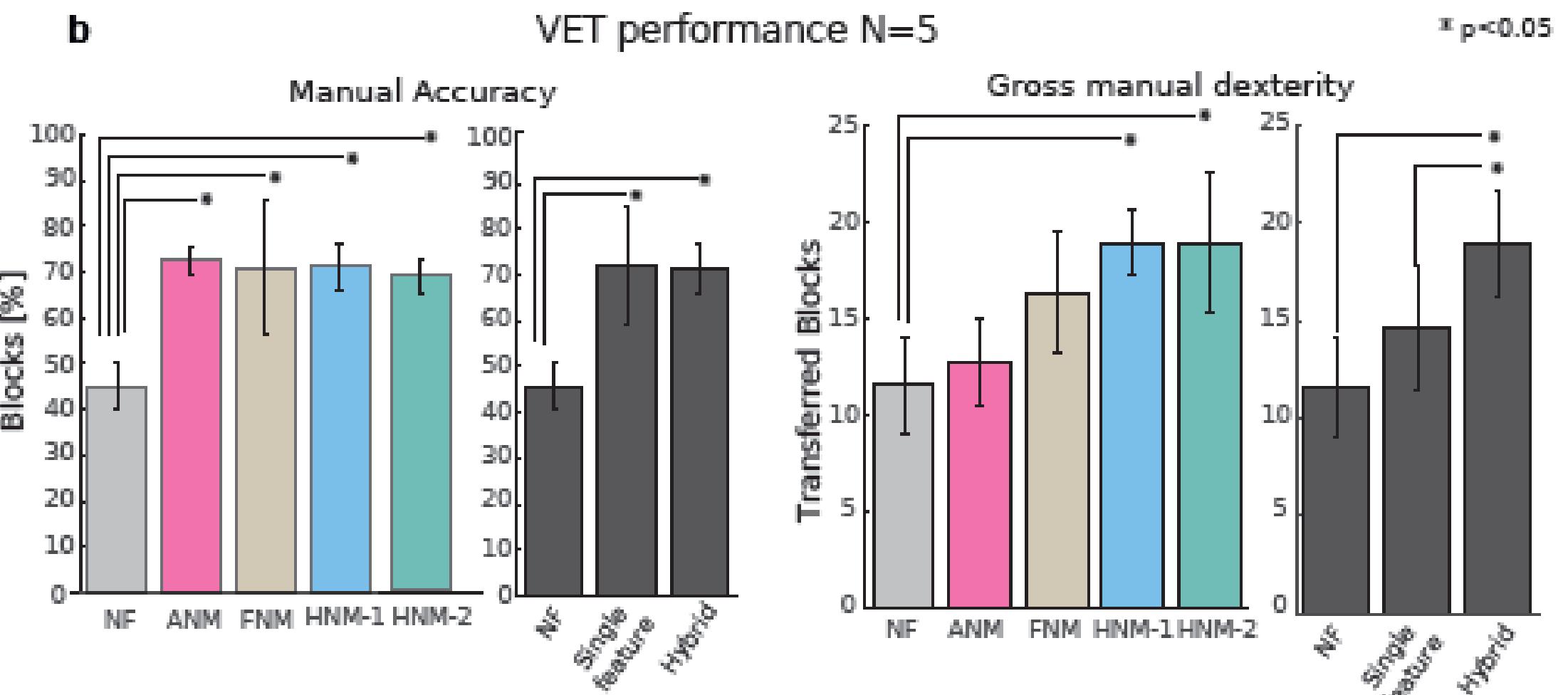
a Setup - Virtual Eggs Test (VET)



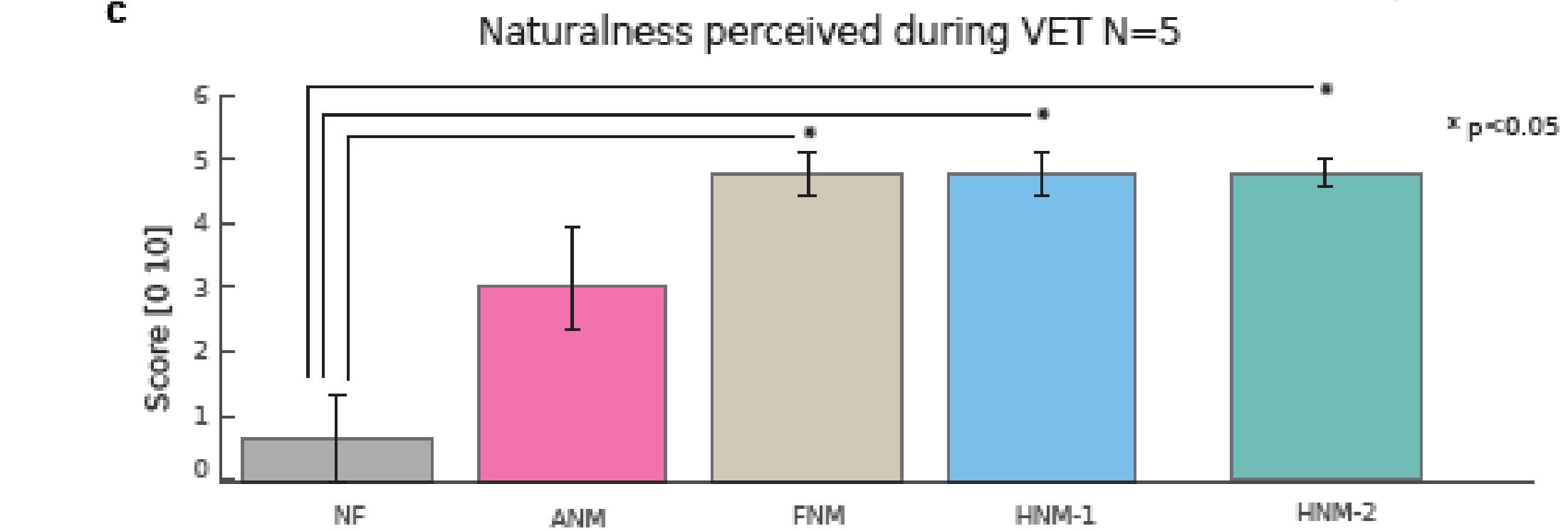
Q1) It seemed like I was grasping a real object

Q2) I felt the intensity of the grasping force applied by the robotic hand on the object

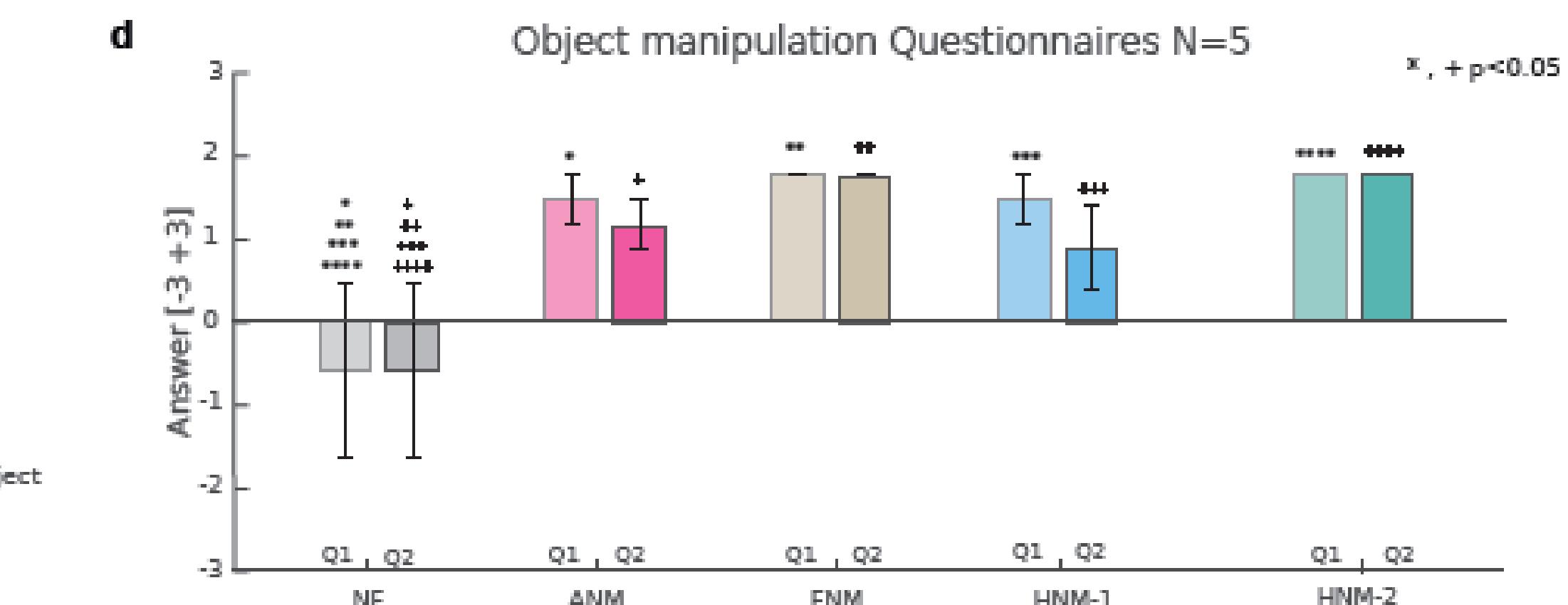
b



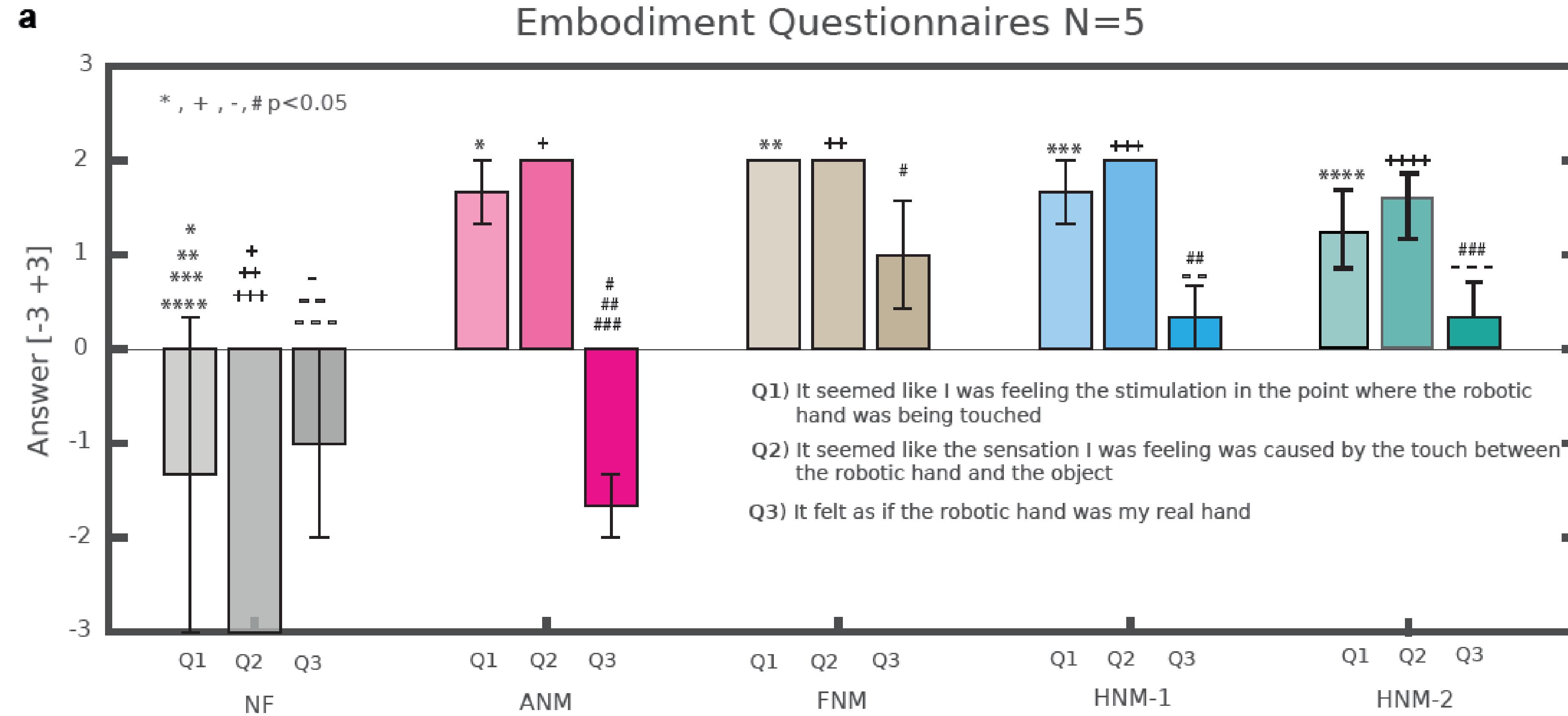
c



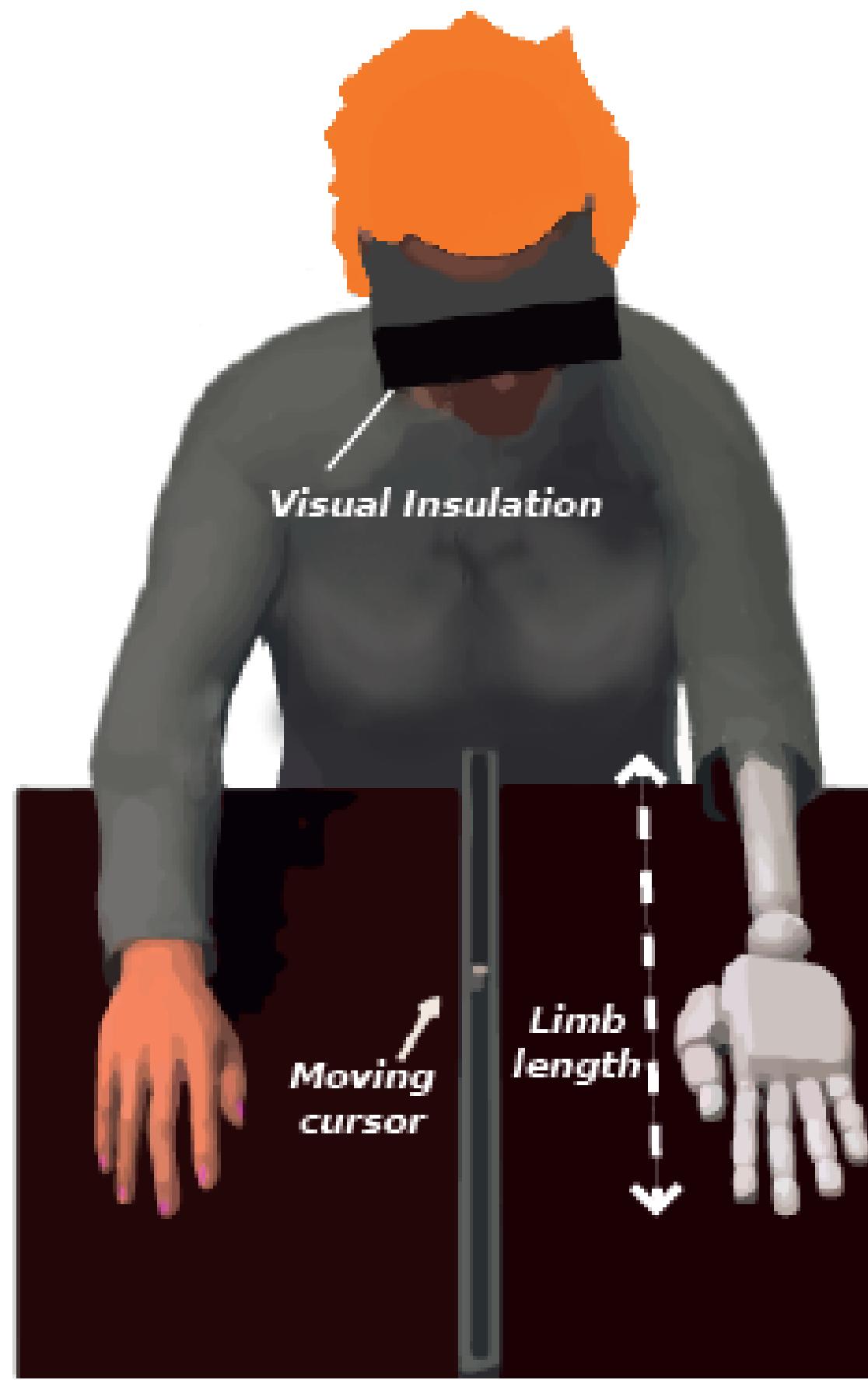
d



Biomimetic encoding strategy

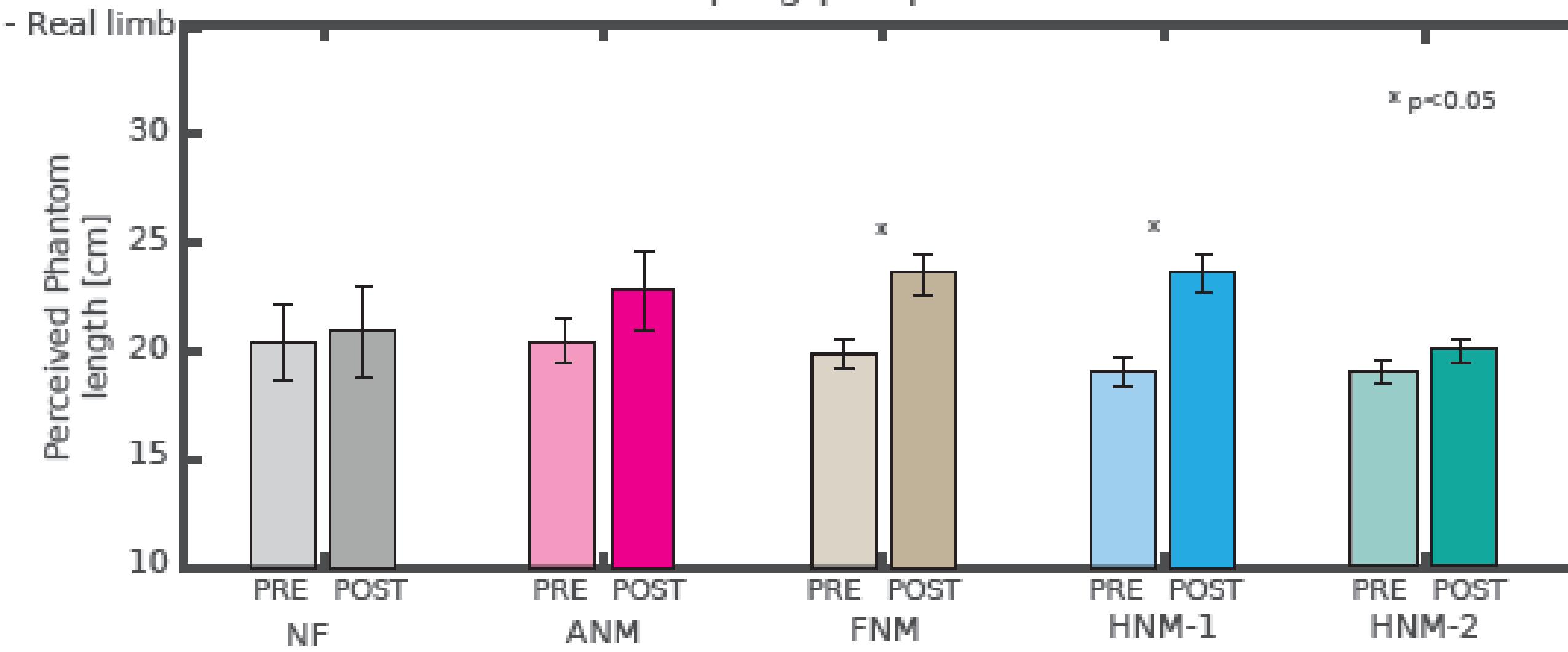


a Telescoping task setup

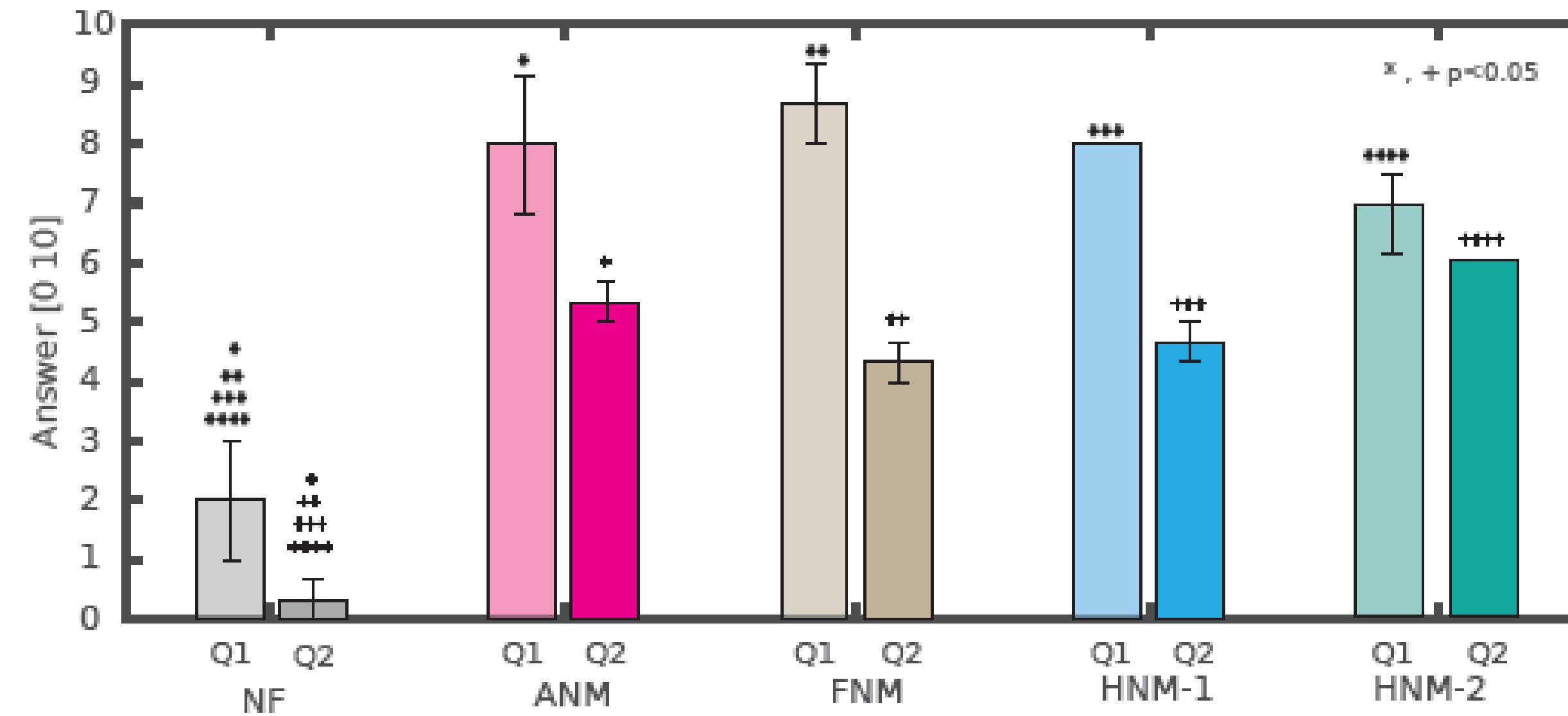


Q1) It seemed like the phantom hand had changed orientation as the robotic hand
Q2) I felt my phantom arm longer

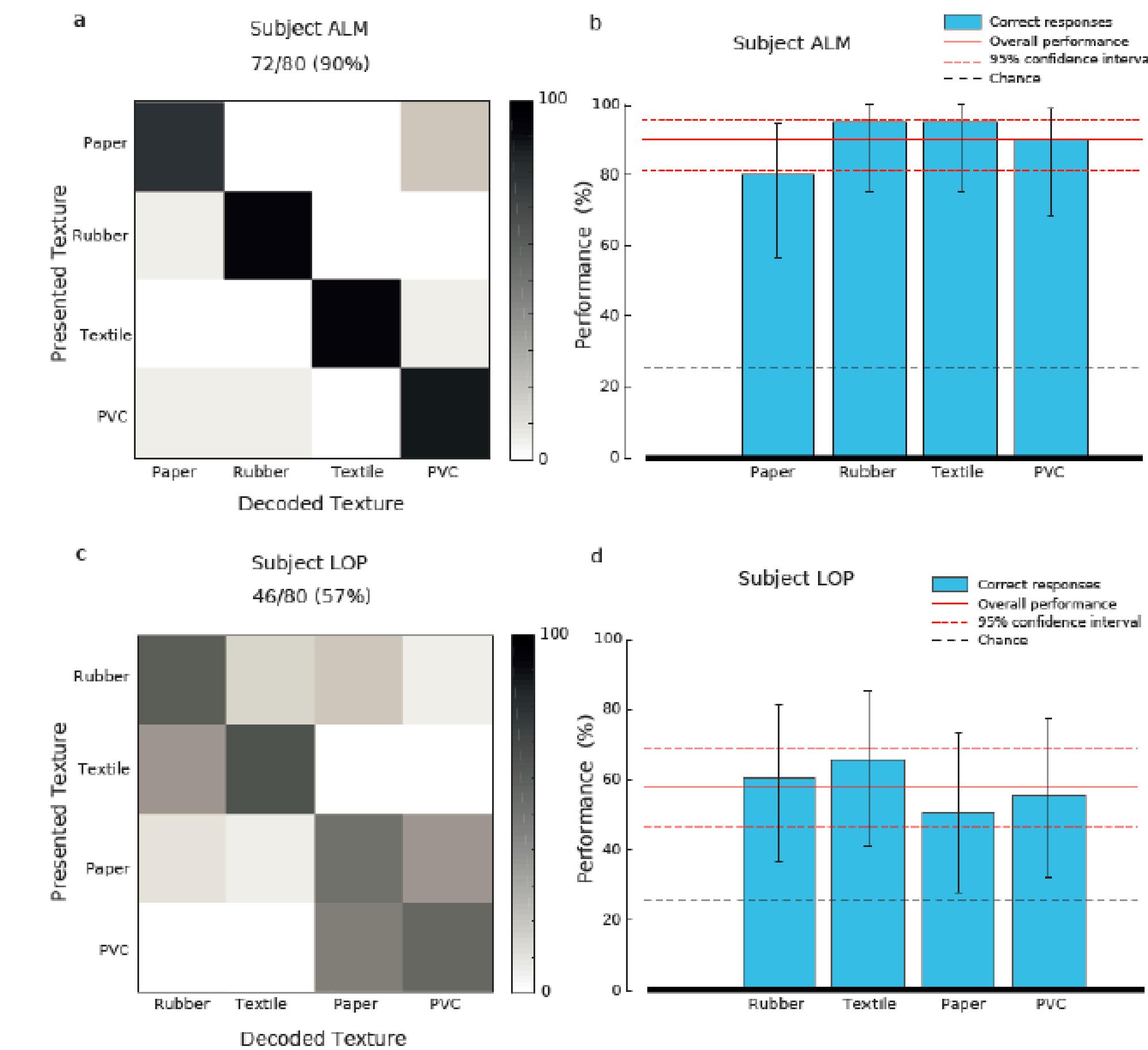
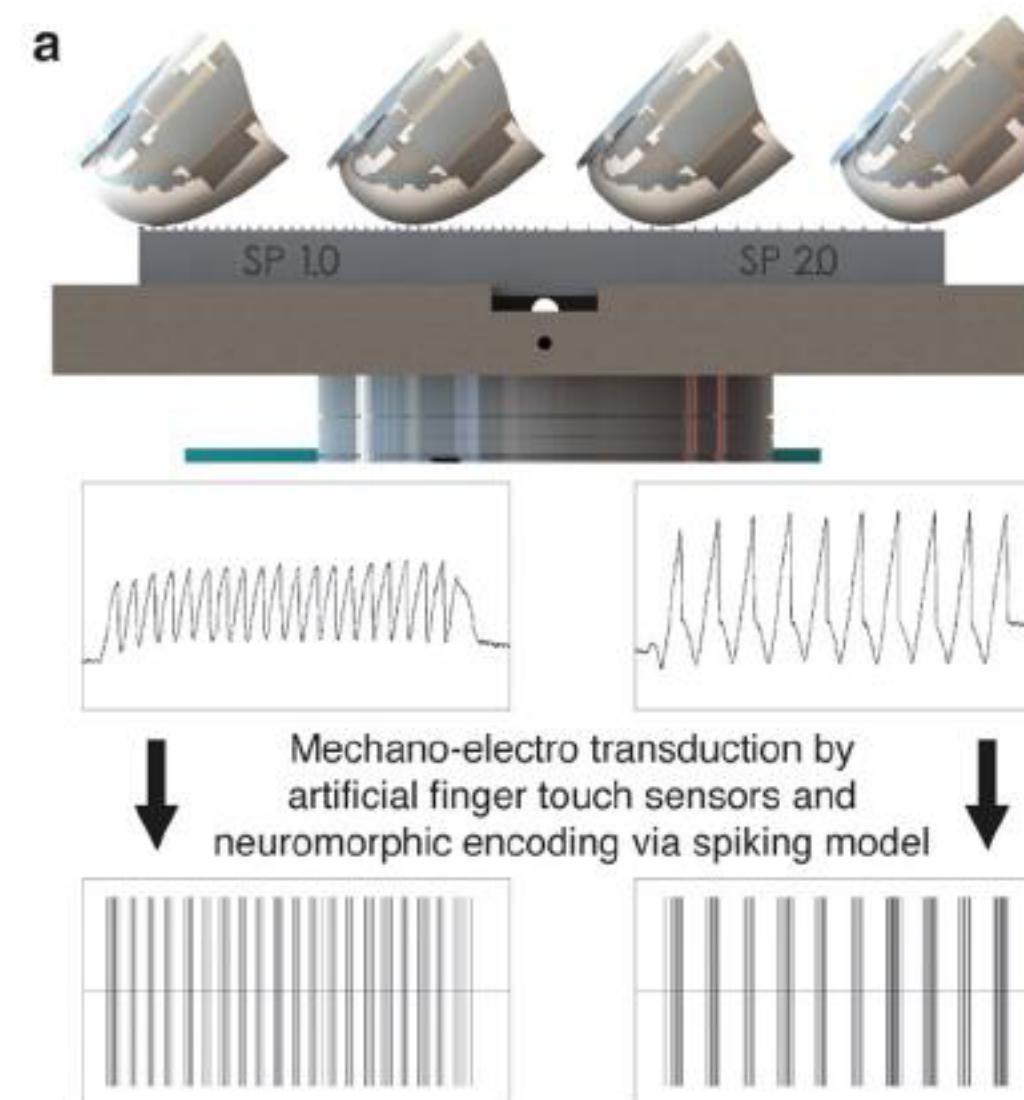
b Telescoping pre-post VET N=5



c Phantom limb dimension perceptions Questionnaires N=5

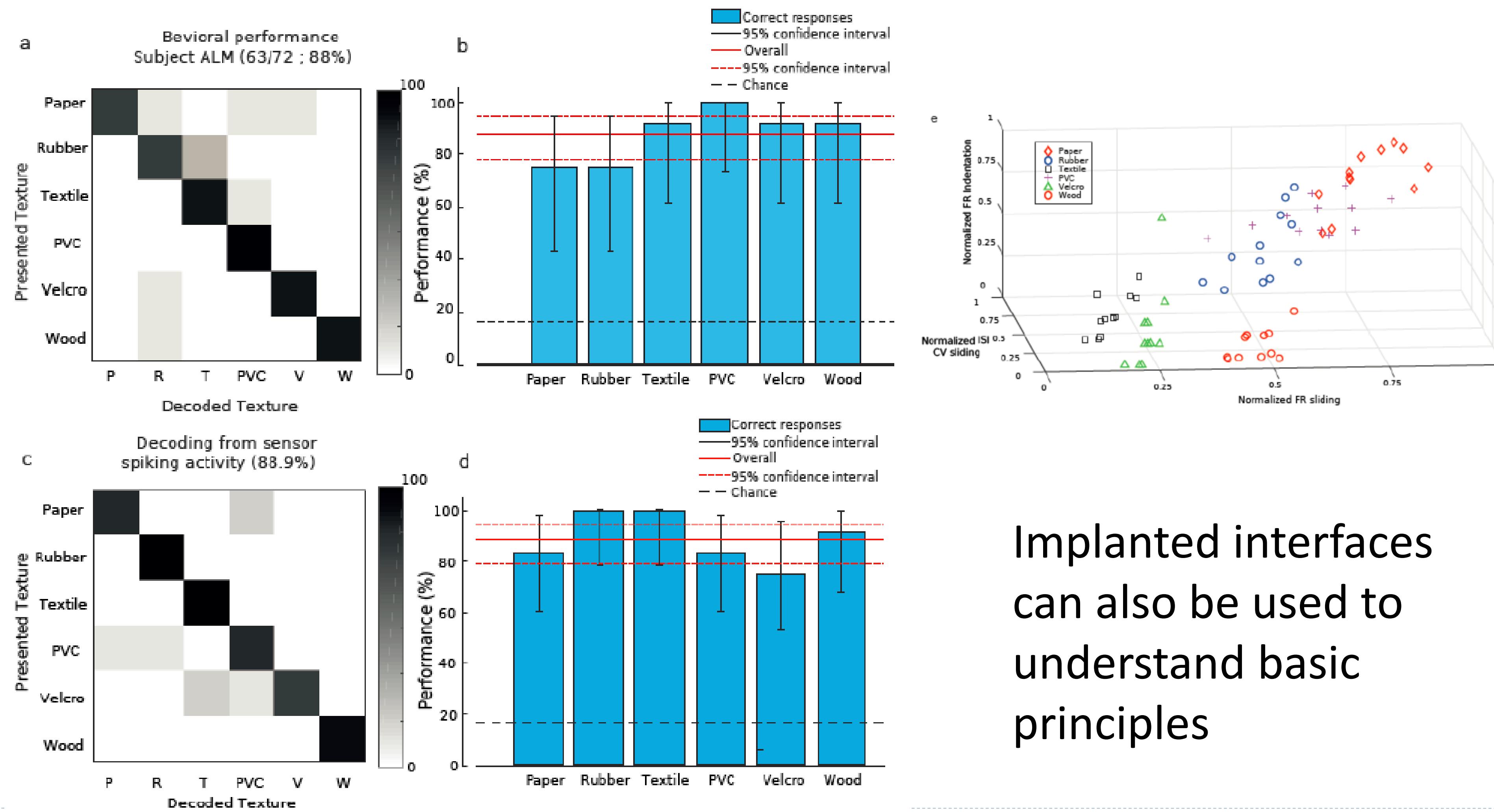


Restoring perception of real textures



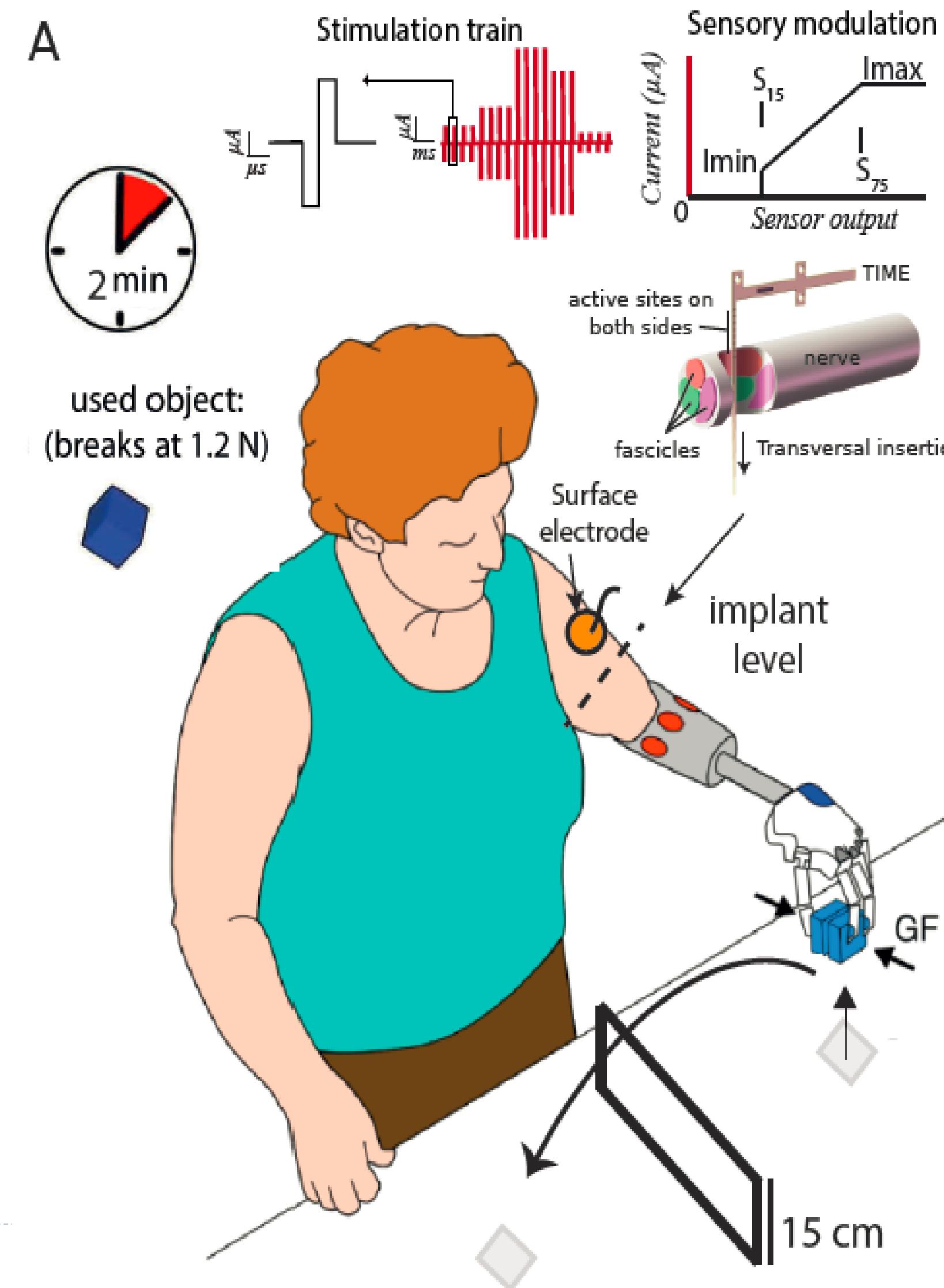
Ondo et al., eLIFE, 2016
Mazzoni et al., Sci Rep, 2019

Restoring perception of real textures



Implanted interfaces
can also be used to
understand basic
principles

Effects of cognitive load



B Induced sensations & stimulation parameters

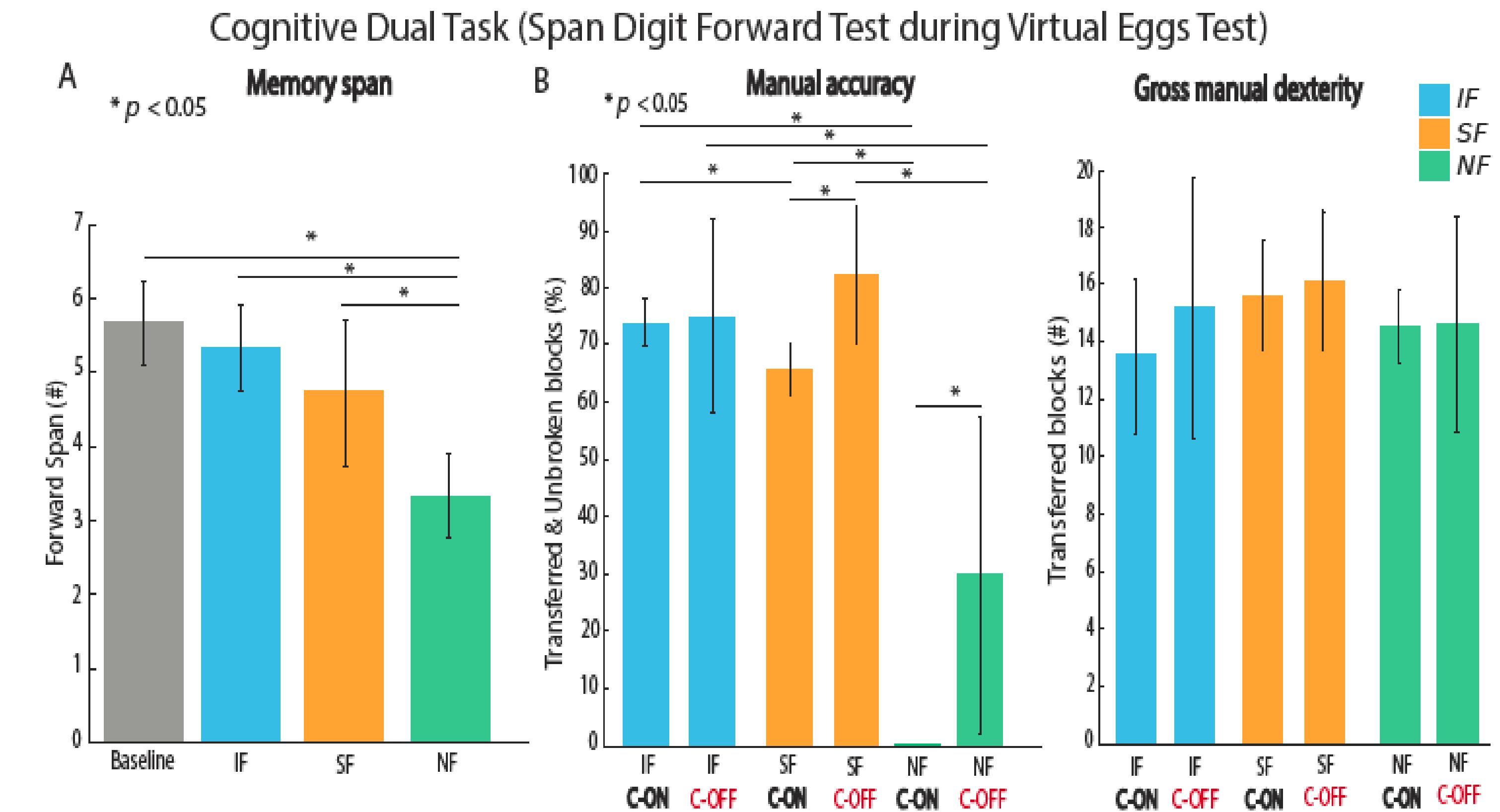
Intraneuronal sensory Feedback (IF)

sensation type	vibration
sensation intensity	$S_{\min} = 1, S_{\max} = 8$
electrode position	proximal part of ulnar nerve above elbow
amplitude	$A_{\min} = 200 \mu A, A_{\max} = 300 \mu A$
pulse-width	$80 \mu s$
frequency	50 Hz

Superficial sensory Feedback (SF)

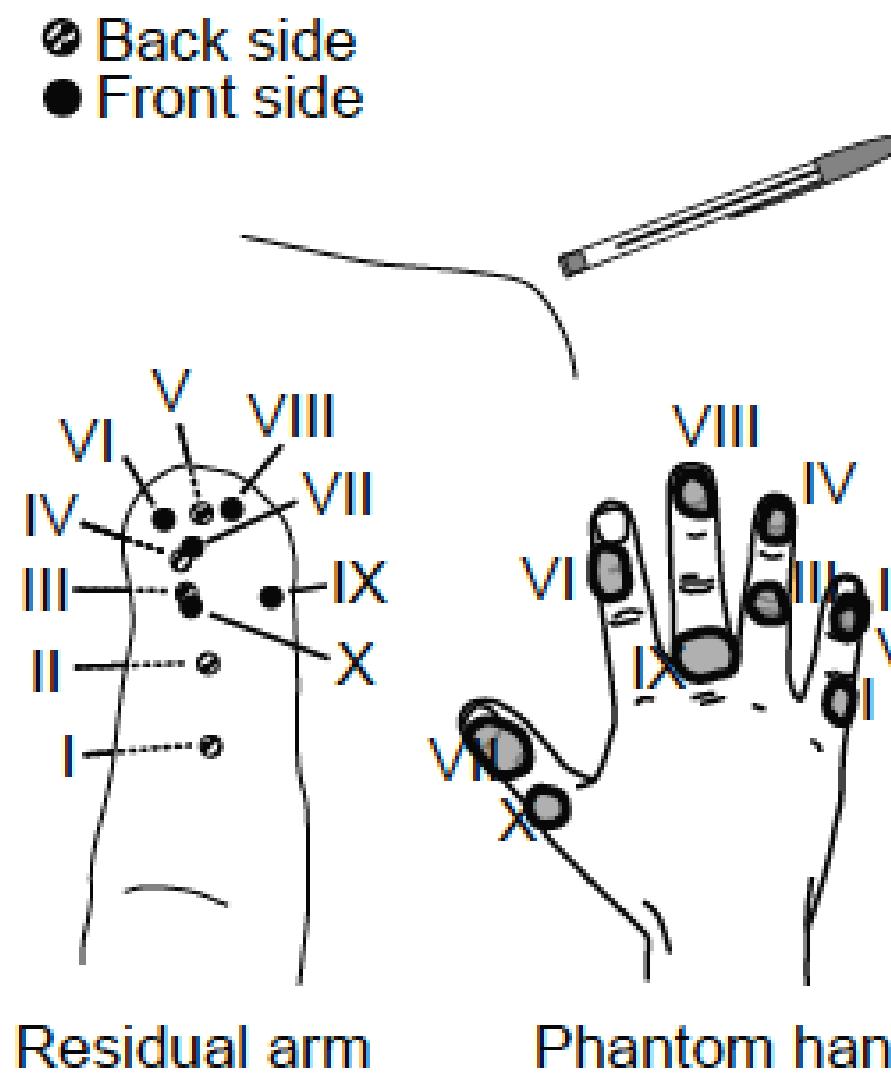
sensation type	electricity
sensation intensity	$S_{\min} = 1, S_{\max} = 8$
electrode position	on the skin of the left arm
amplitude	$A_{\min} = 100 \mu A, A_{\max} = 500 \mu A$
pulse-width	$200 \mu s$
frequency	50 Hz

Effects of cognitive load

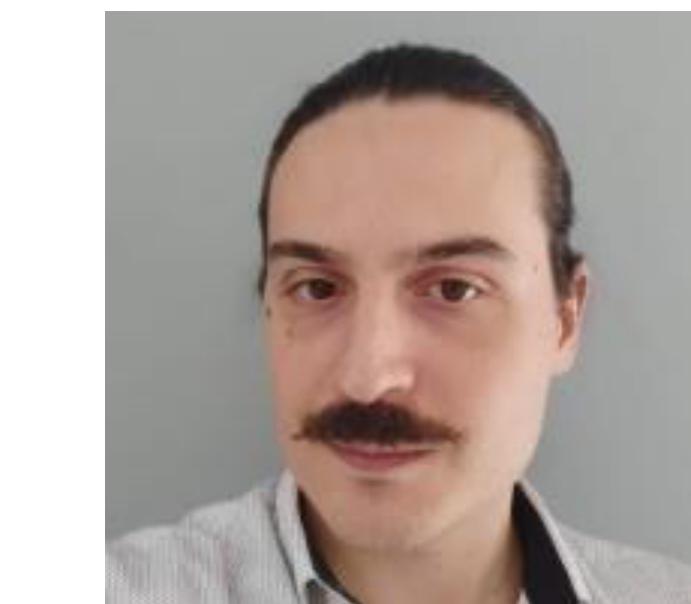


Bidirectional neurocontrolled hand prostheses

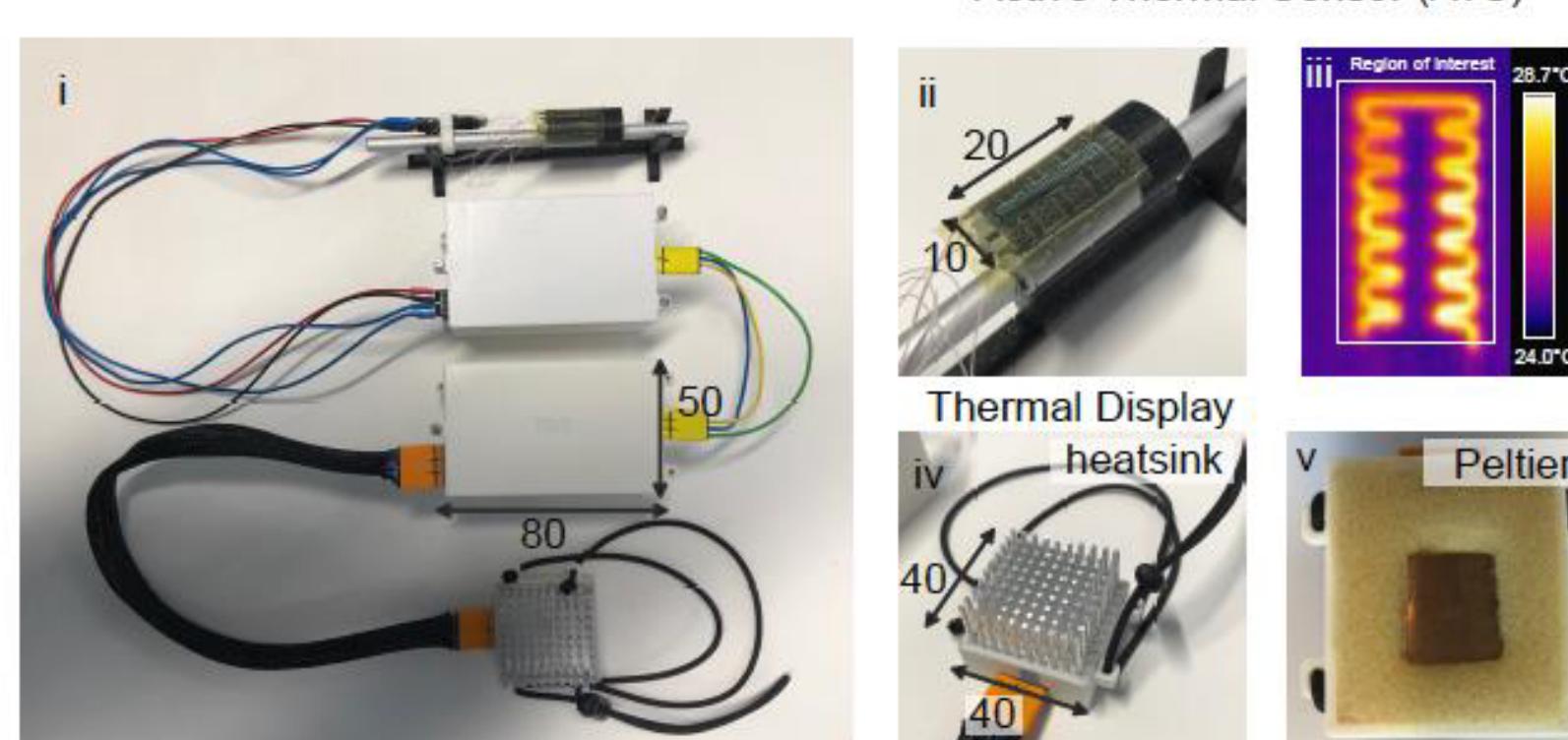
“Multimodal” sensory feedback



Residual arm Phantom hand

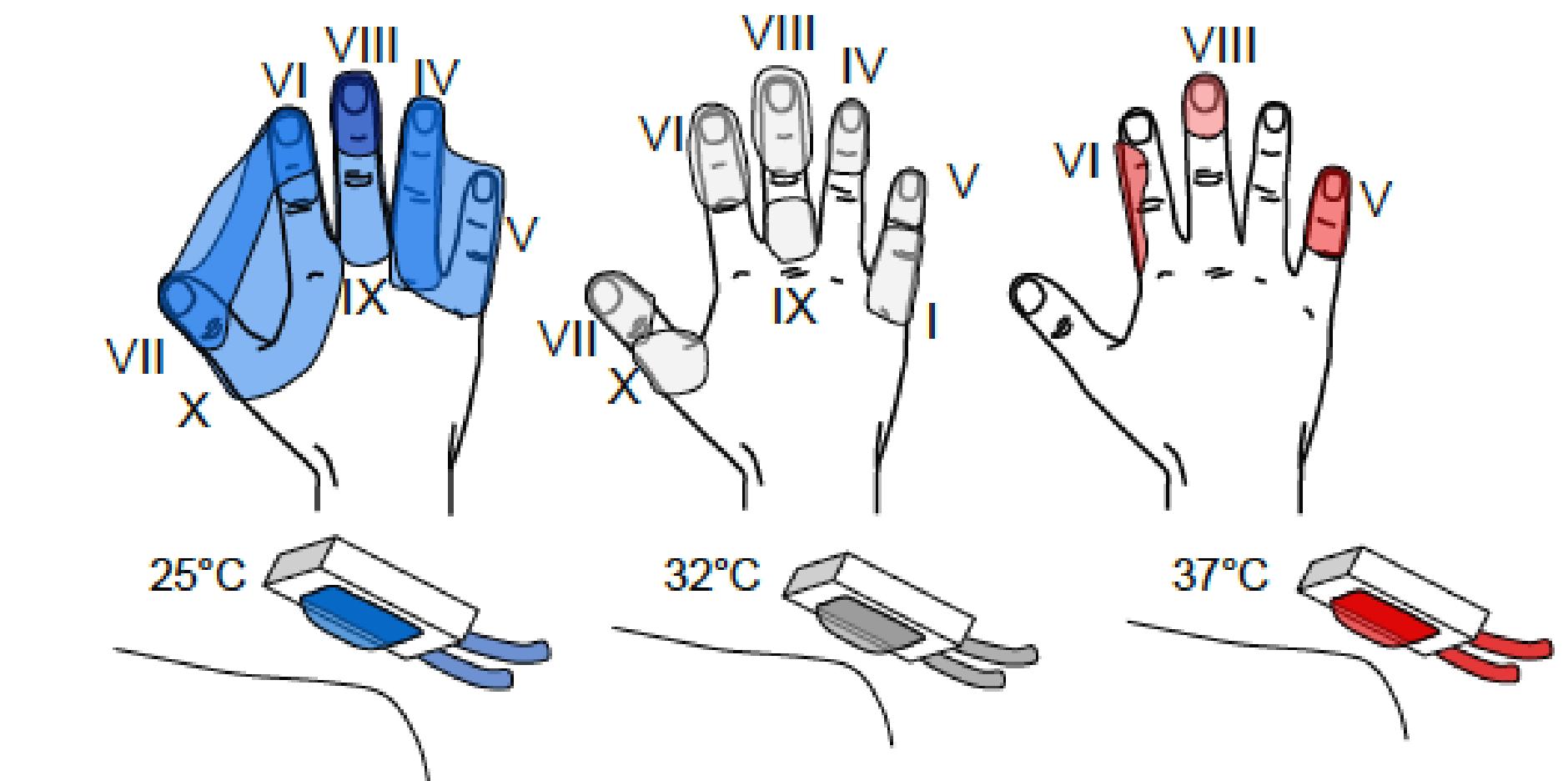


S.Micera

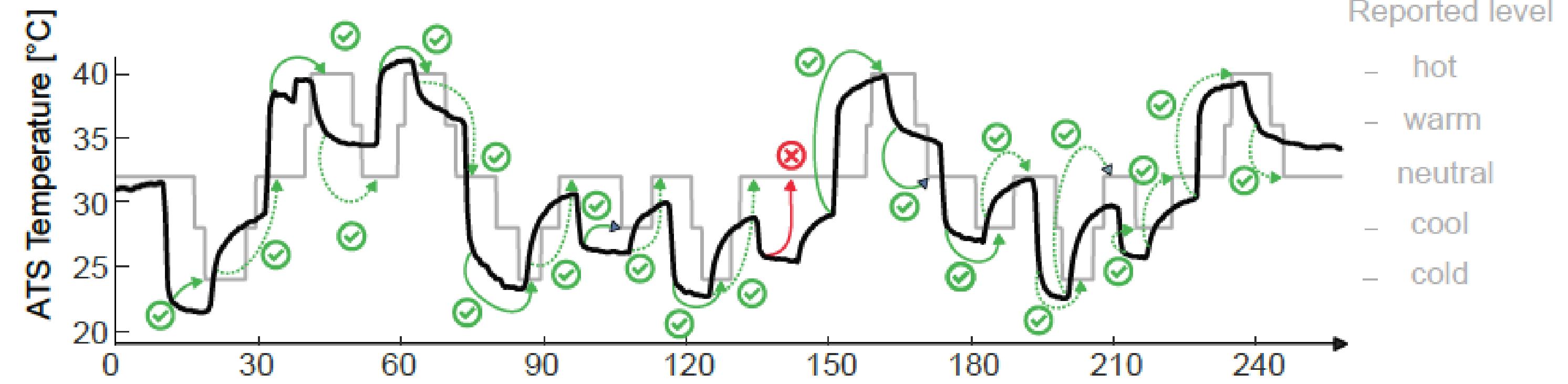


i4

Iberite et al., Science, 2023

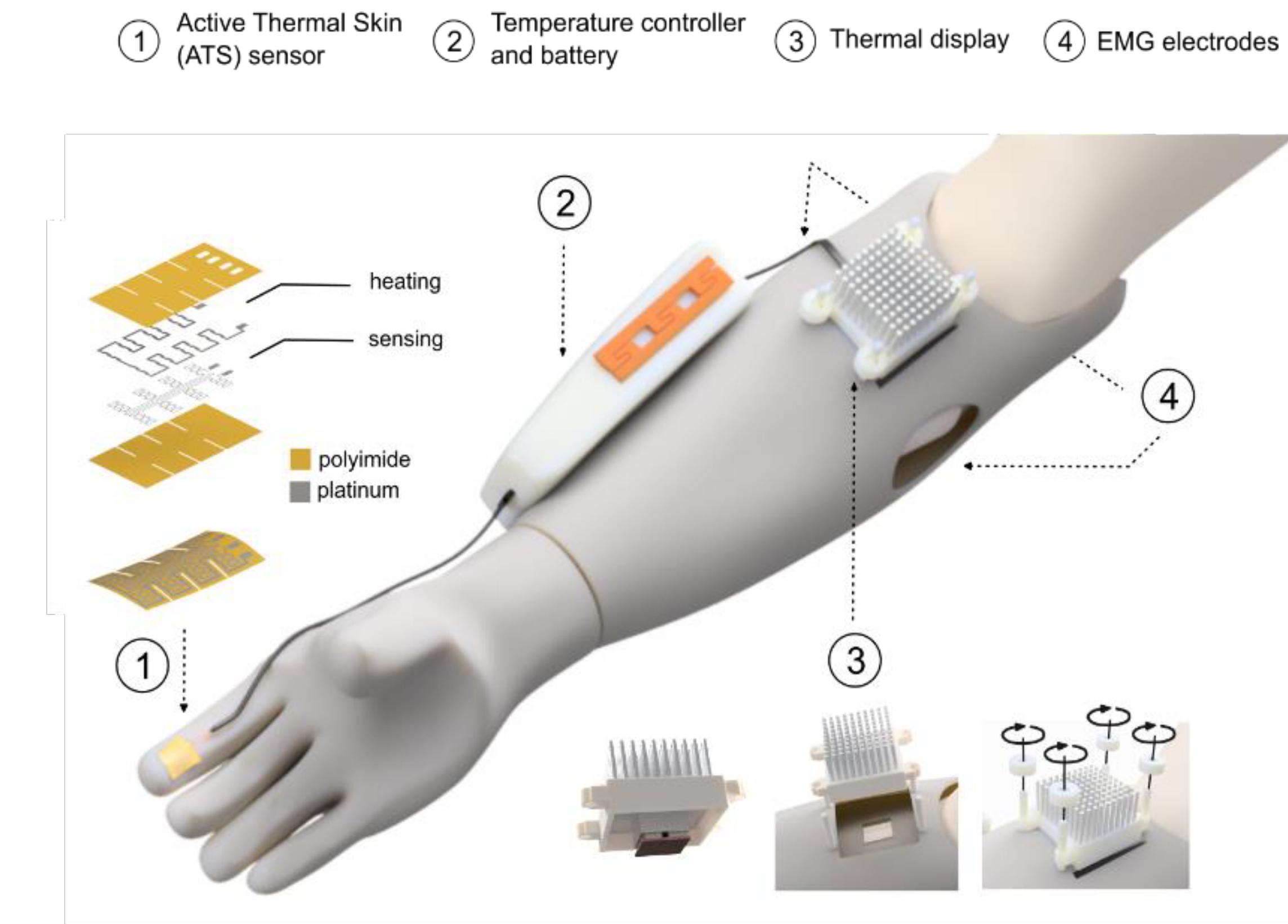
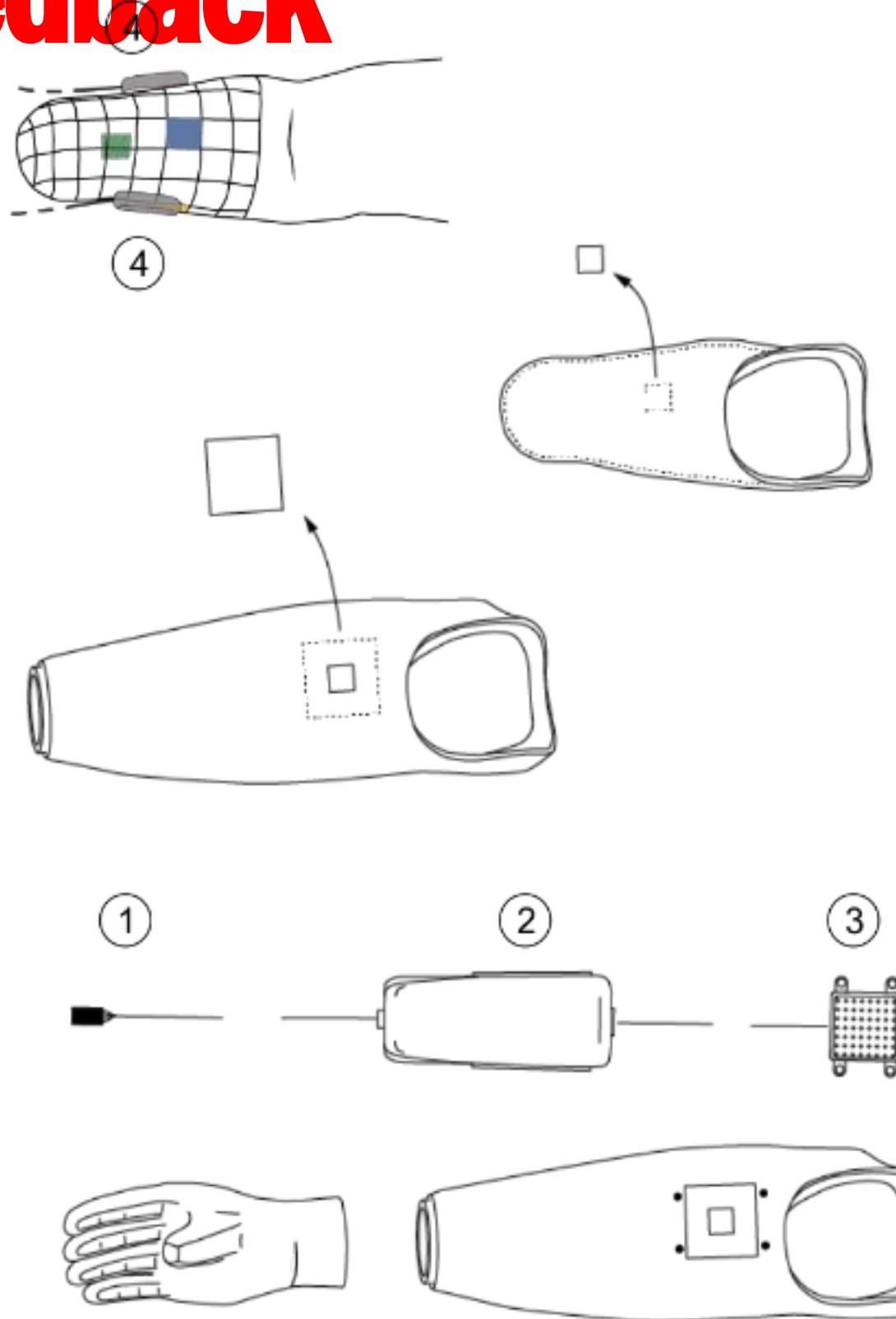
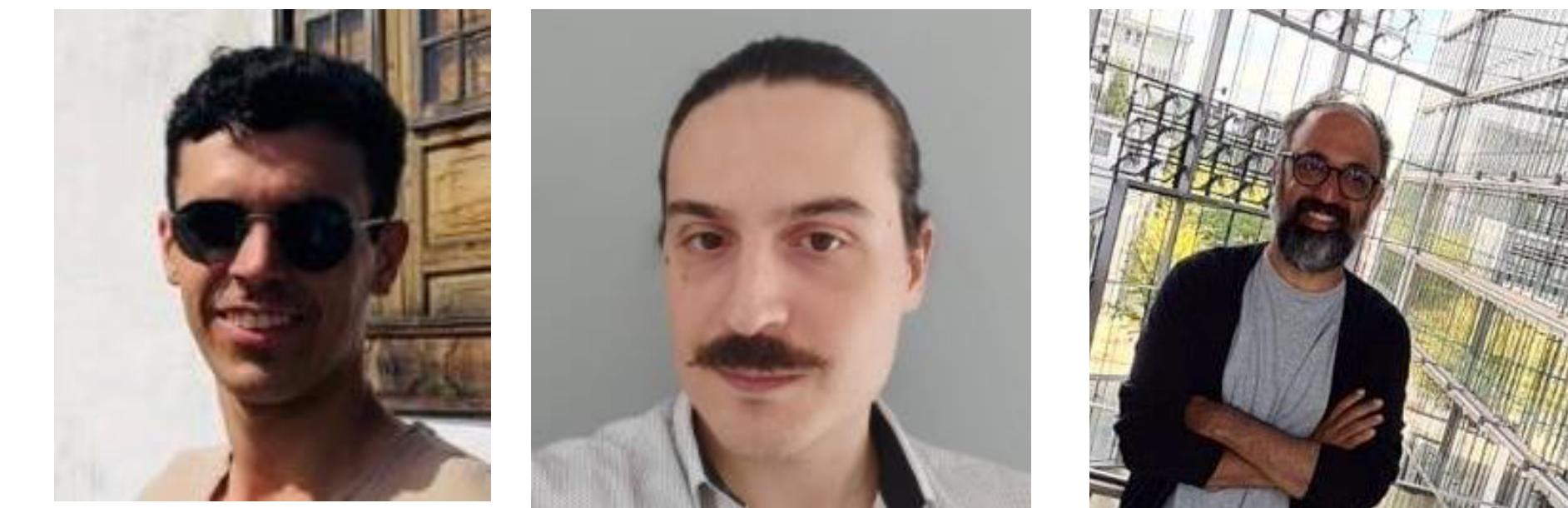


B. Real-time Thermal perception (P12)



Bidirectional neurocontrolled hand prostheses

“Multimodal” sensory feedback



Bidirectional neurocontrolled hand prostheses

“Multimodal” sensory feedback

A. Temperature discrimination



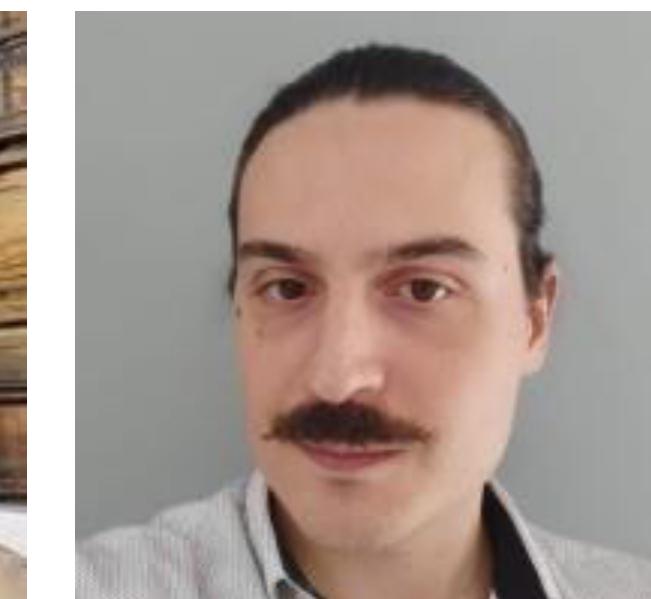
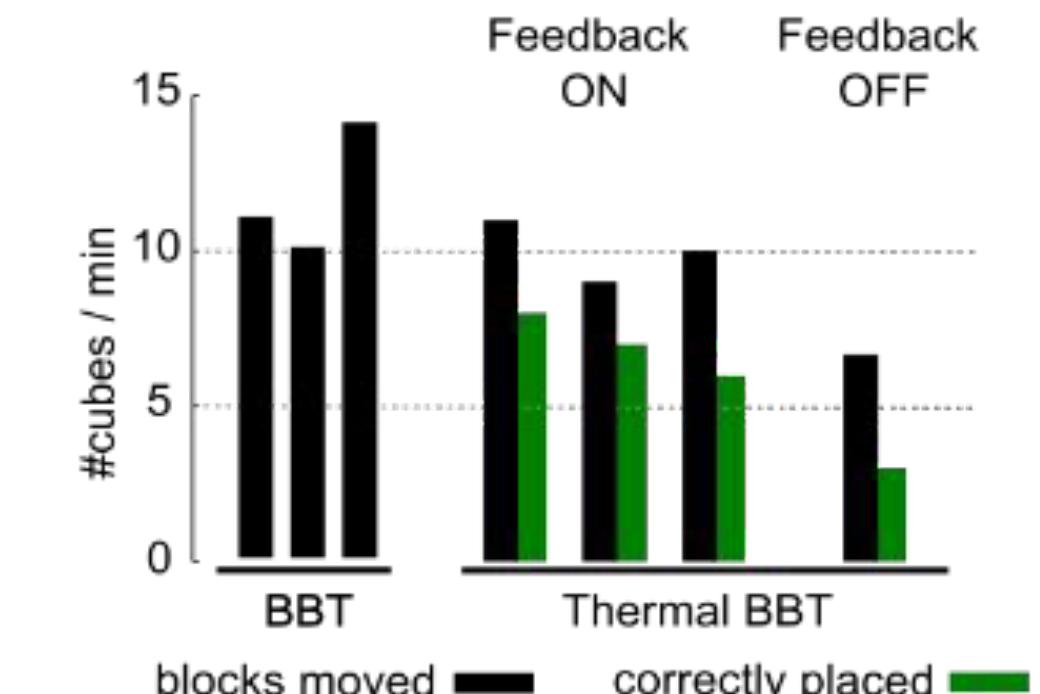
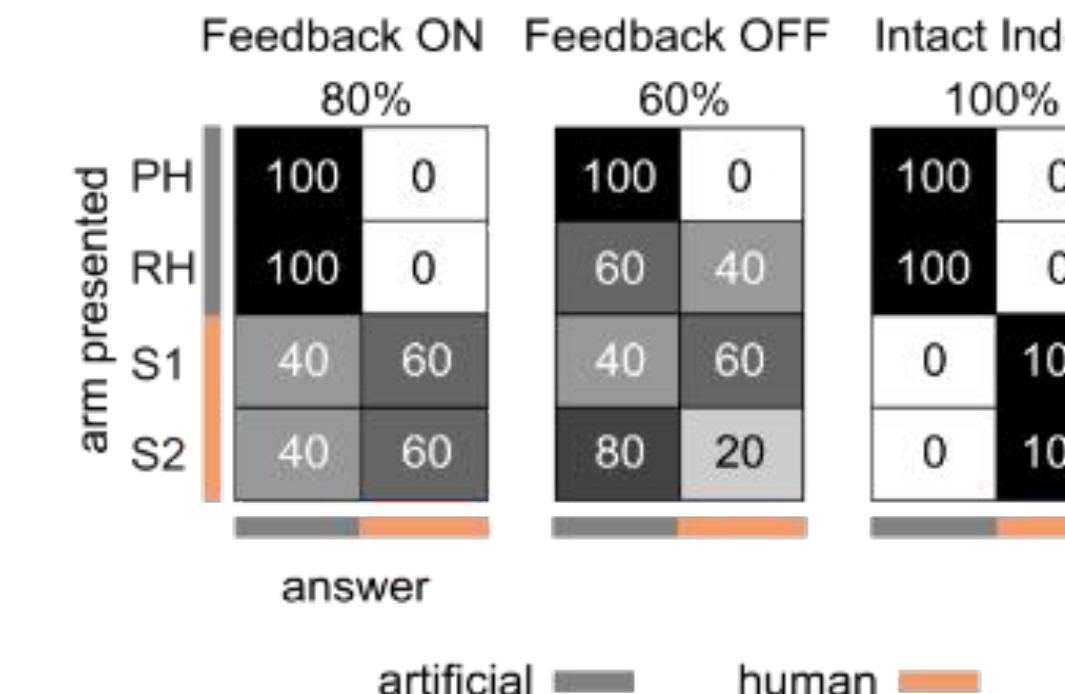
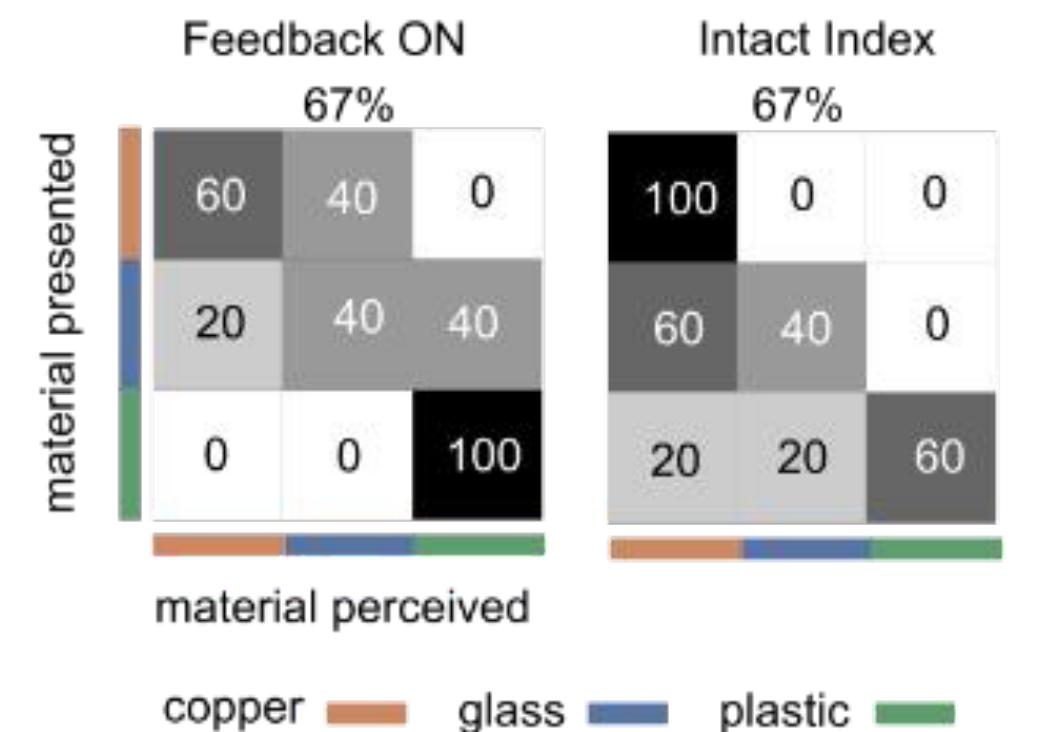
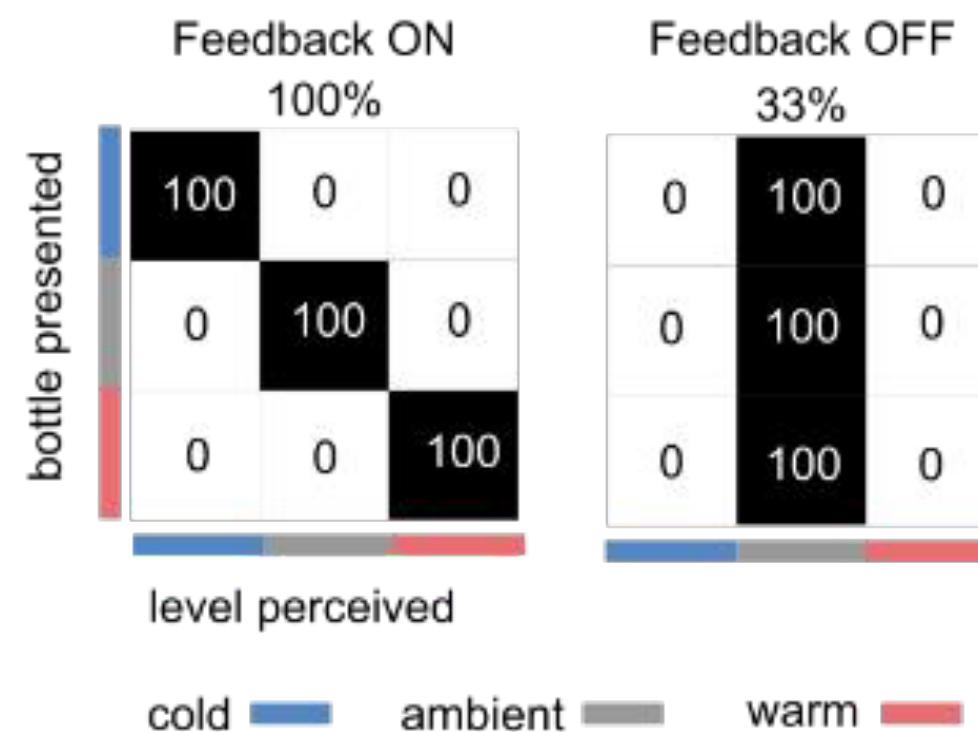
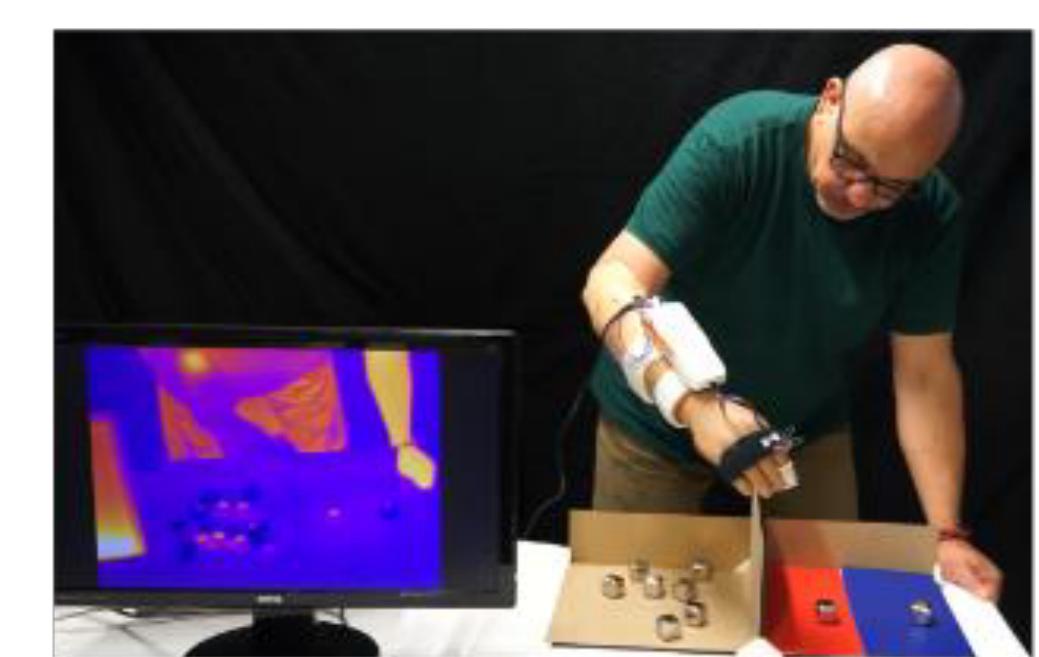
B. Material discrimination



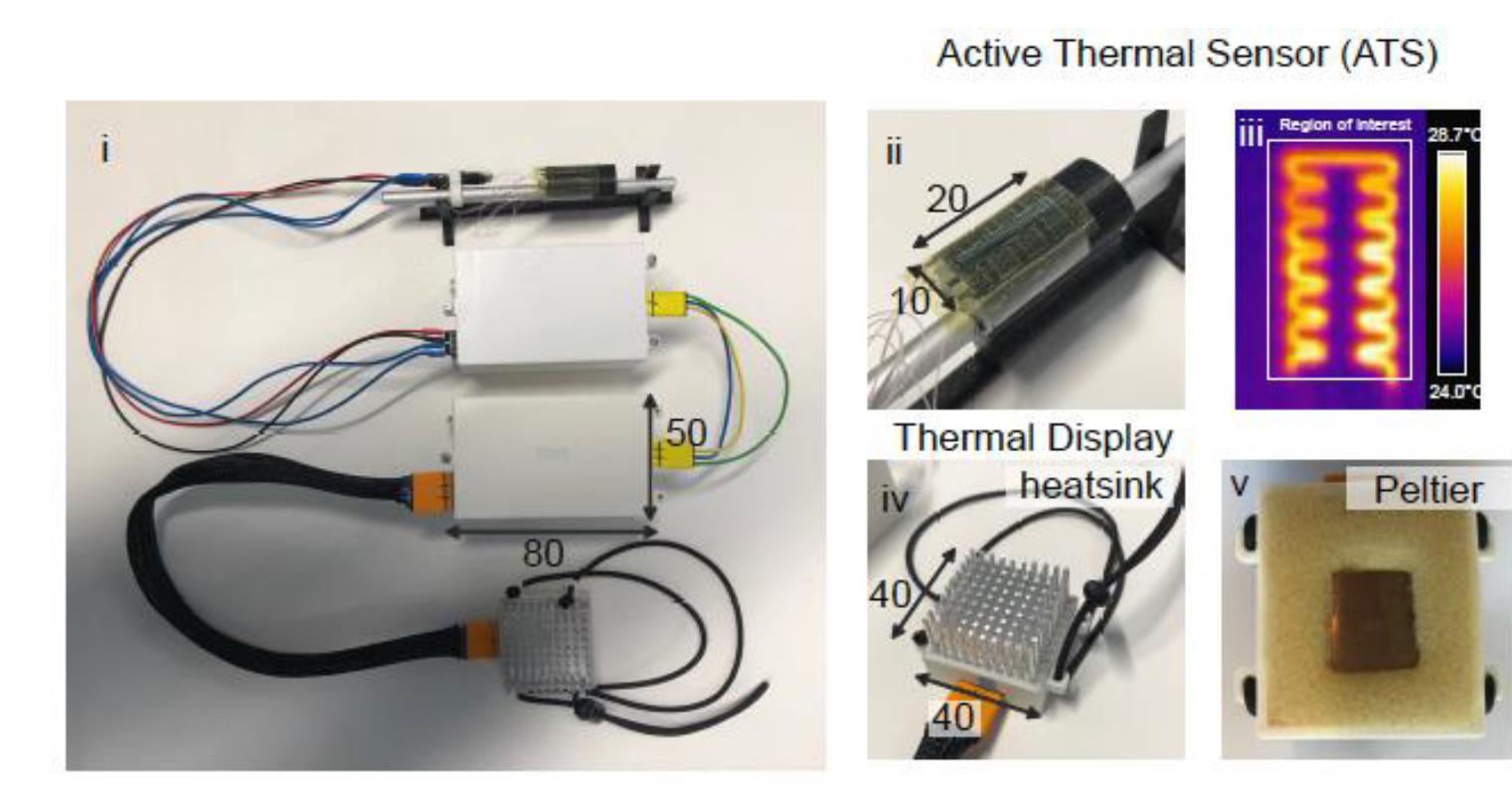
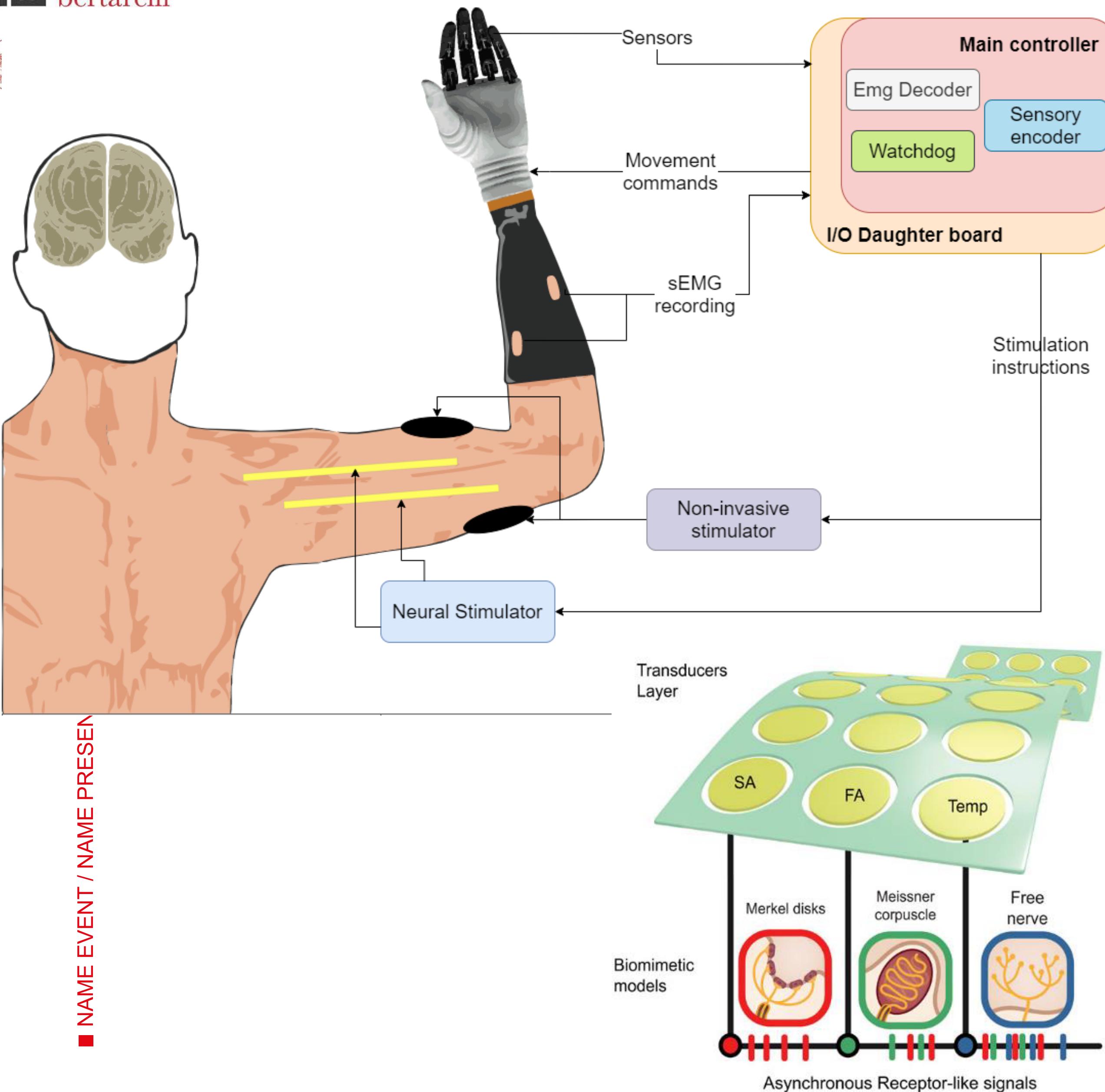
C. Bodily contact



D. Thermal Box and Blocks Test



NEXT STEP – Going chronic at home



Bidirectional neurocontrolled leg prostheses



Above the knee

Below the knee

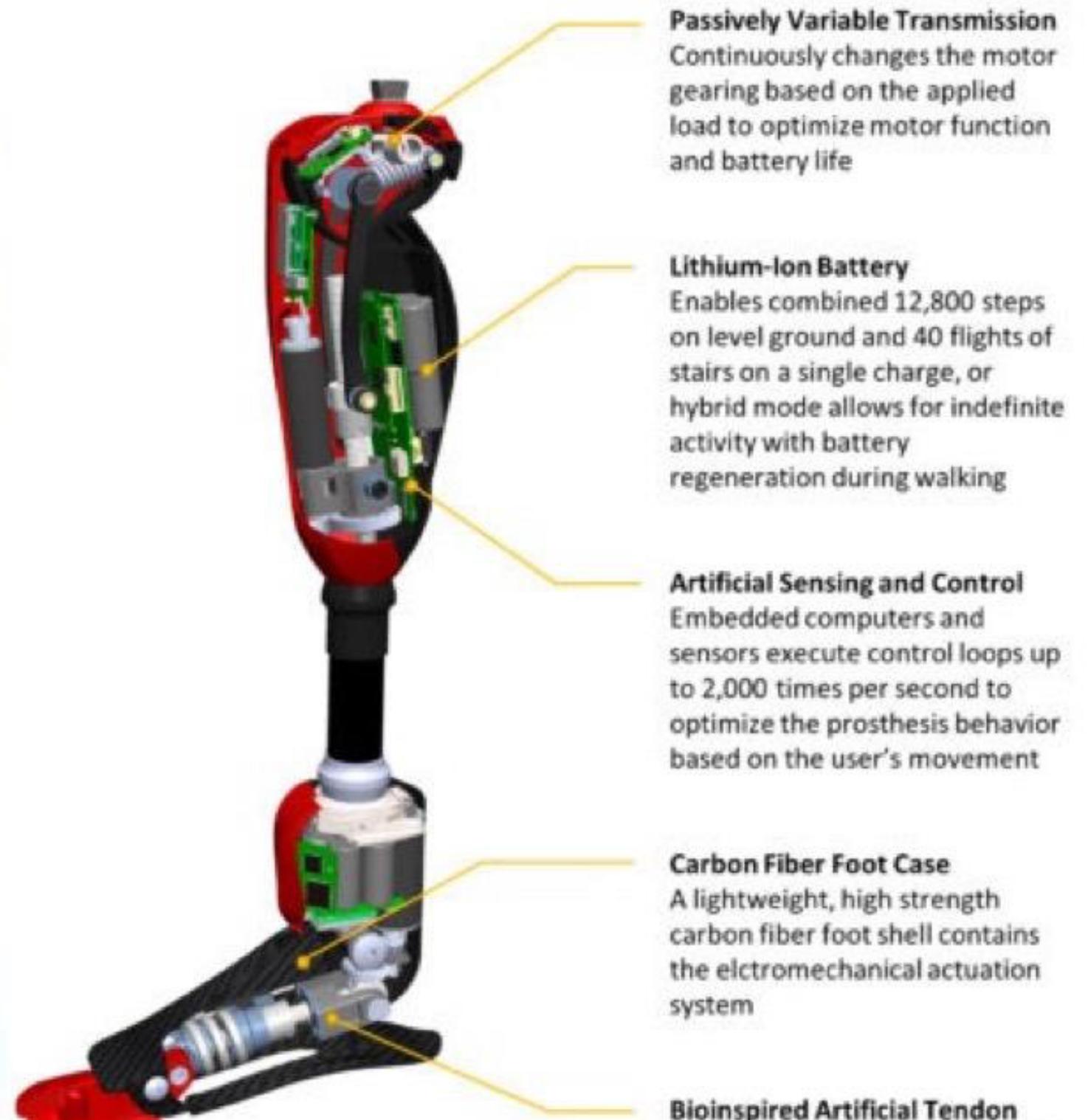
Leg Prosthetics

Utah Bionic Leg

Powered Knee Module
 Weight: 1.6 kg
 Range of Motion: 120 deg
 Max Torque: 150 Nm
 Max Speed: 500 deg/s
 Build Height: 255mm

Standard Connection
 Allows adjustment of prosthesis build height and ankle inversion/eversion to patient using standard prosthetic components

Powered Ankle-Toe Module
 Weight: 1.6 kg
 Range of Motion - Ankle: 40 deg
 Range of Motion - Toe: 45 deg
 Max Torque: 150 Nm
 Max Speed: 350 deg/s
 Build Height: 165 mm



Passively Variable Transmission
 Continuously changes the motor gearing based on the applied load to optimize motor function and battery life

Lithium-Ion Battery
 Enables combined 12,800 steps on level ground and 40 flights of stairs on a single charge, or hybrid mode allows for indefinite activity with battery regeneration during walking

Artificial Sensing and Control
 Embedded computers and sensors execute control loops up to 2,000 times per second to optimize the prosthesis behavior based on the user's movement

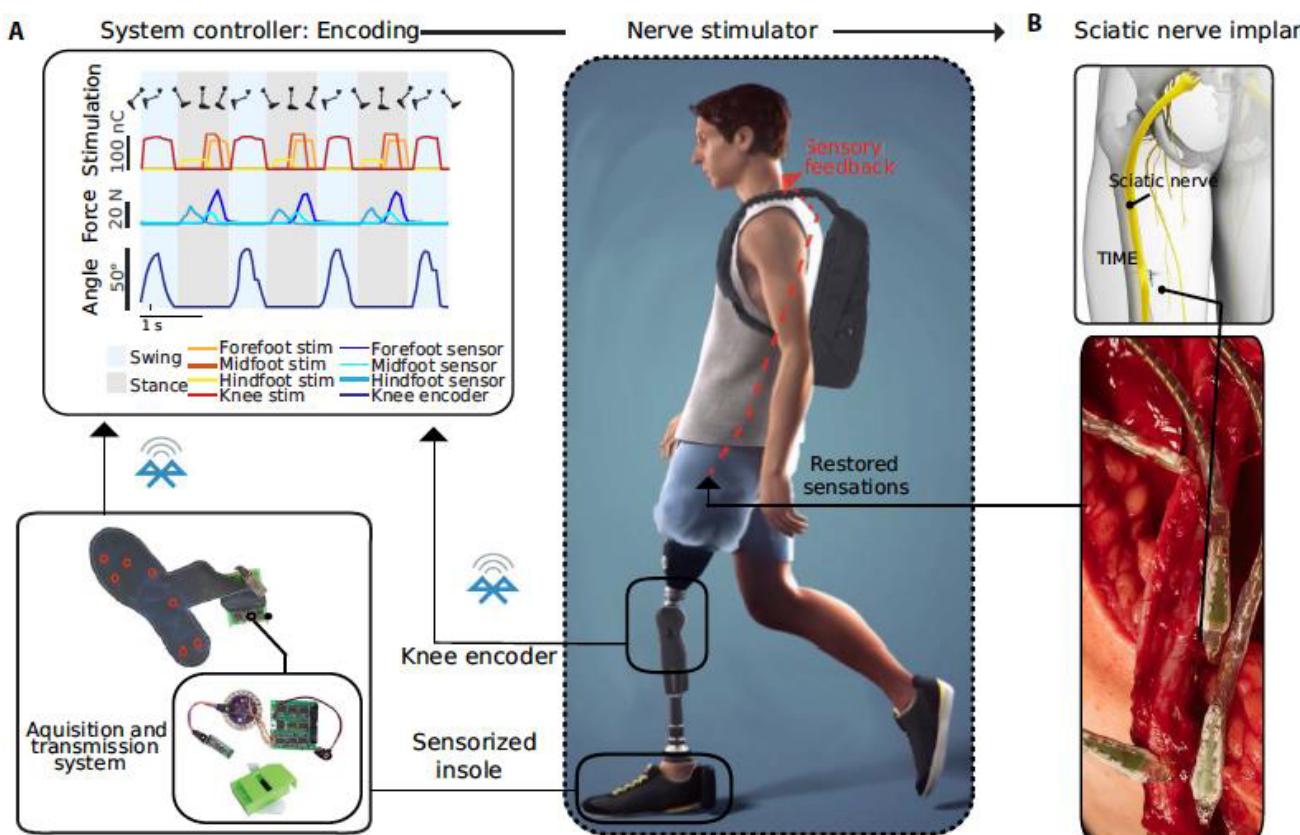
Carbon Fiber Foot Case
 A lightweight, high strength carbon fiber foot shell contains the electromechanical actuation system

Bioinspired Artificial Tendon
 An artificial tendon connects the toe and the ankle joint to allow for biomimetic foot mechanics during walking

Bidirectional neurocontrolled leg prostheses

Sensory feedback

Enhancing functional abilities and cognitive integration
of the lower limb prosthesis



Movie S2:
Neuroprosthesis working
principle and active tasks

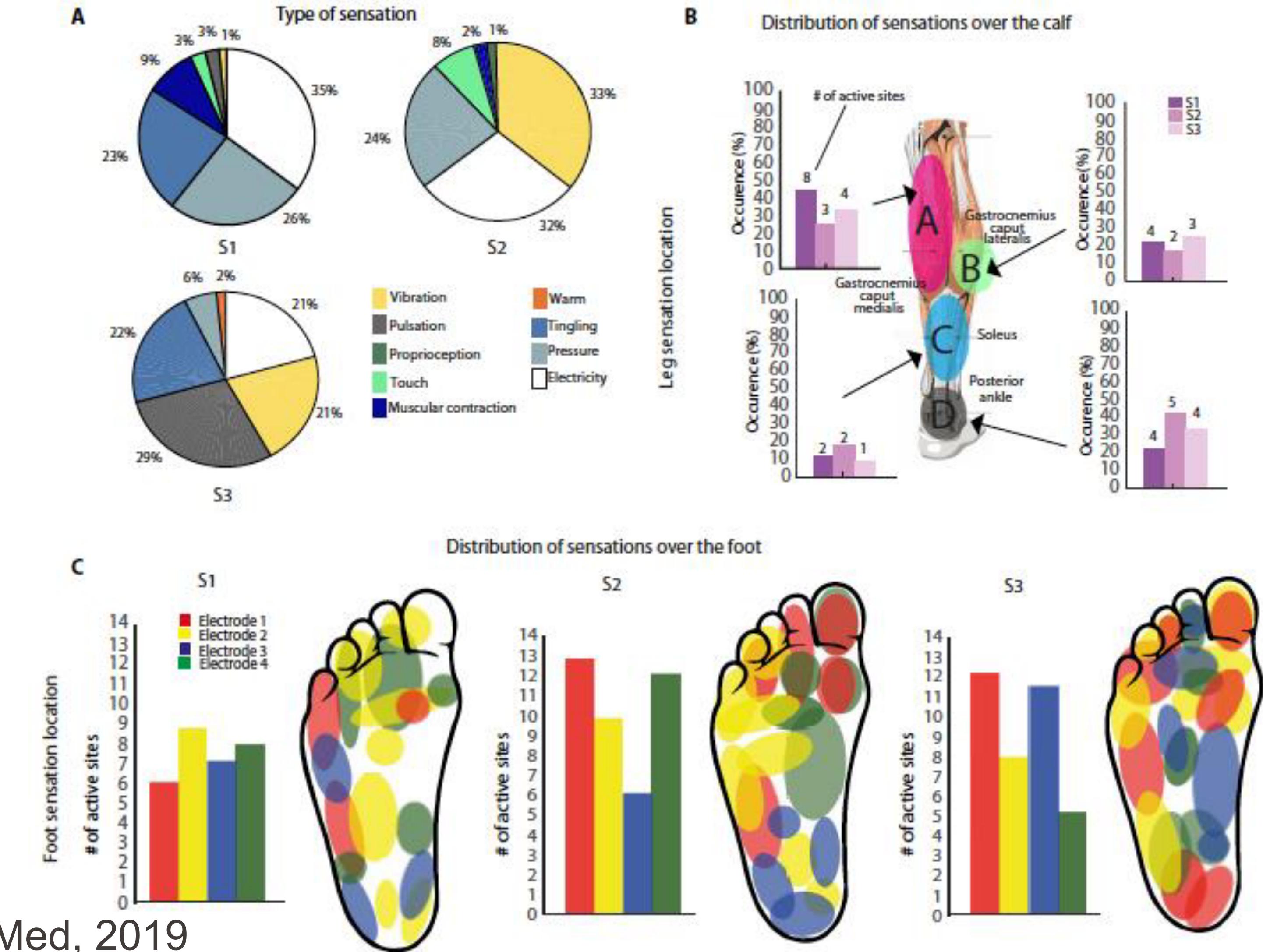
Caution: Investigational device

Bidirectional neurocontrolled leg prostheses

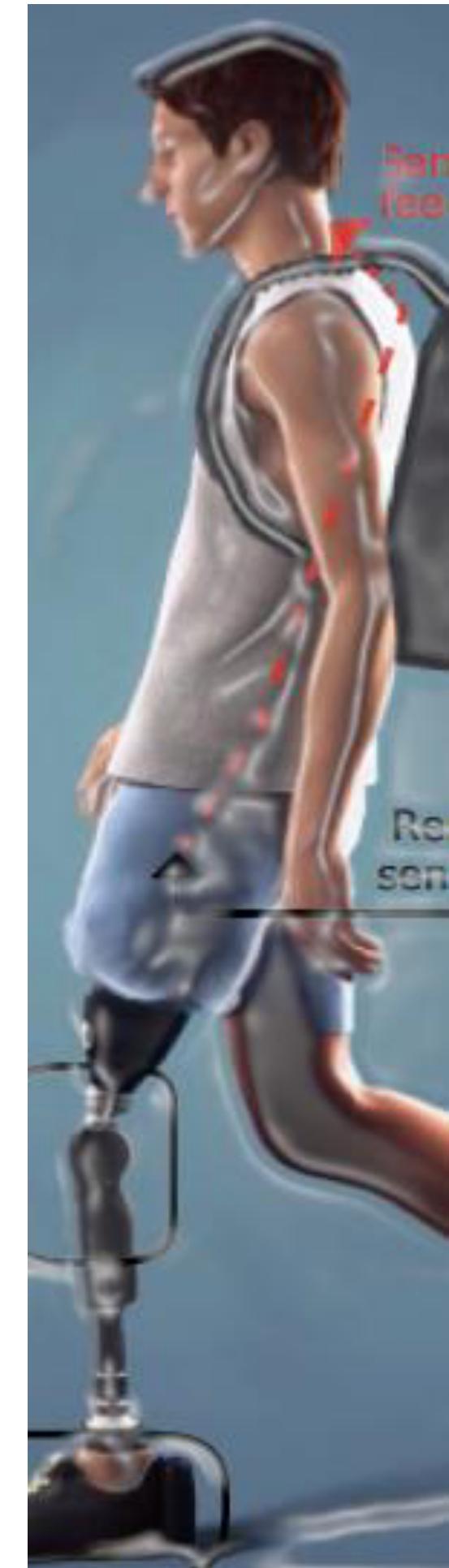
Sensory feedback



i4LIFE – Intraneural stimulation

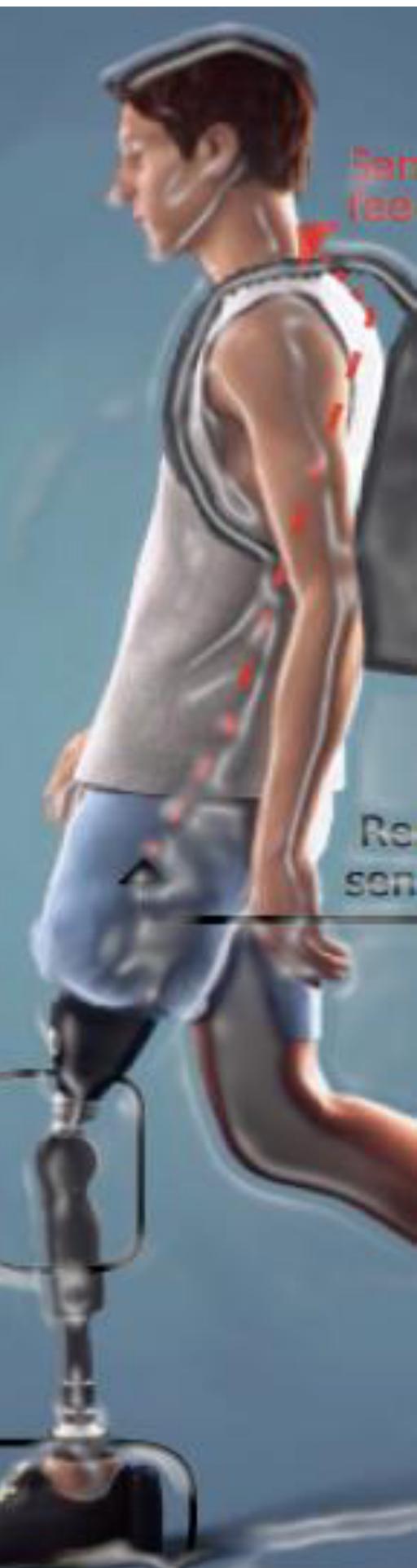
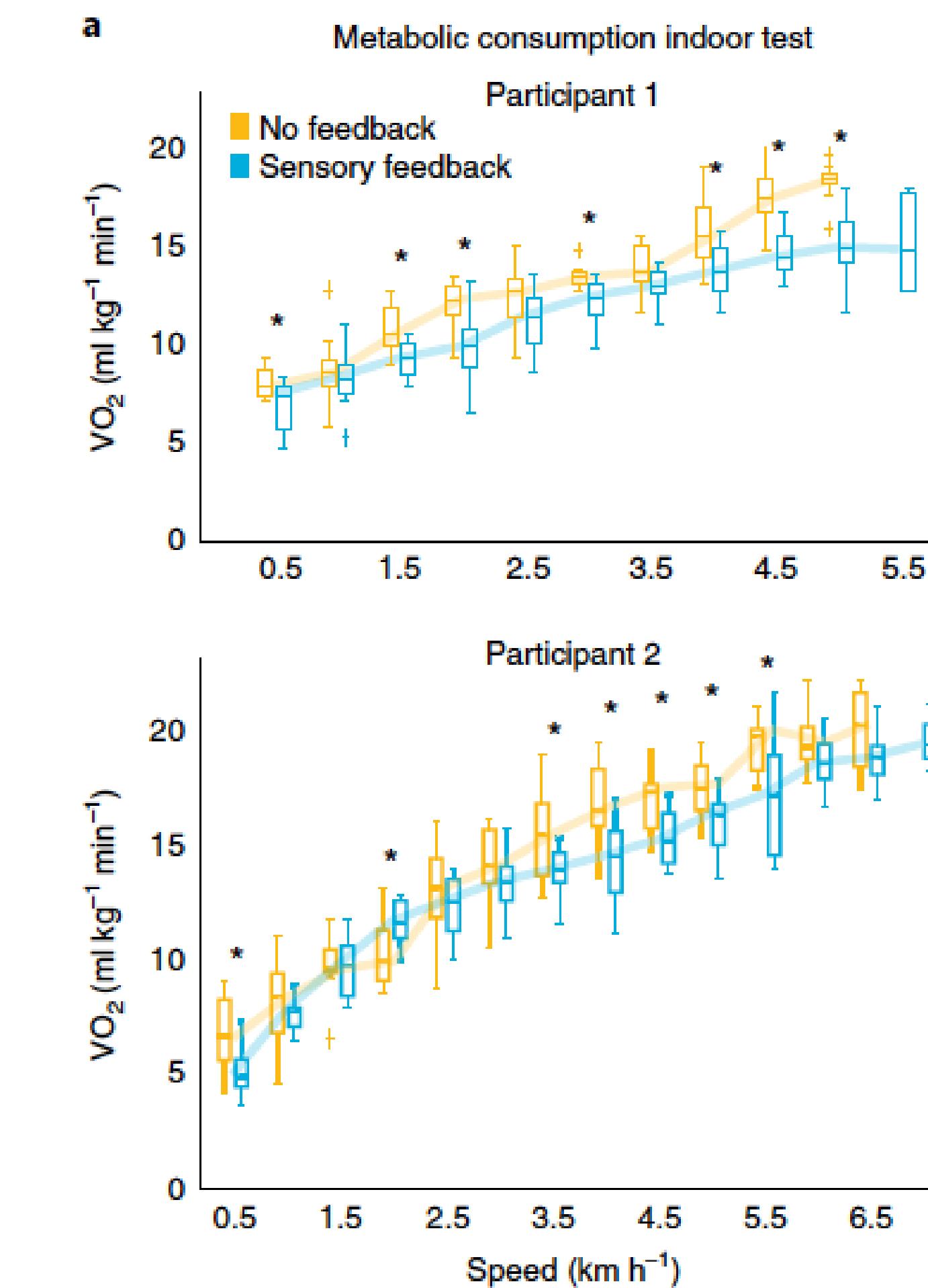
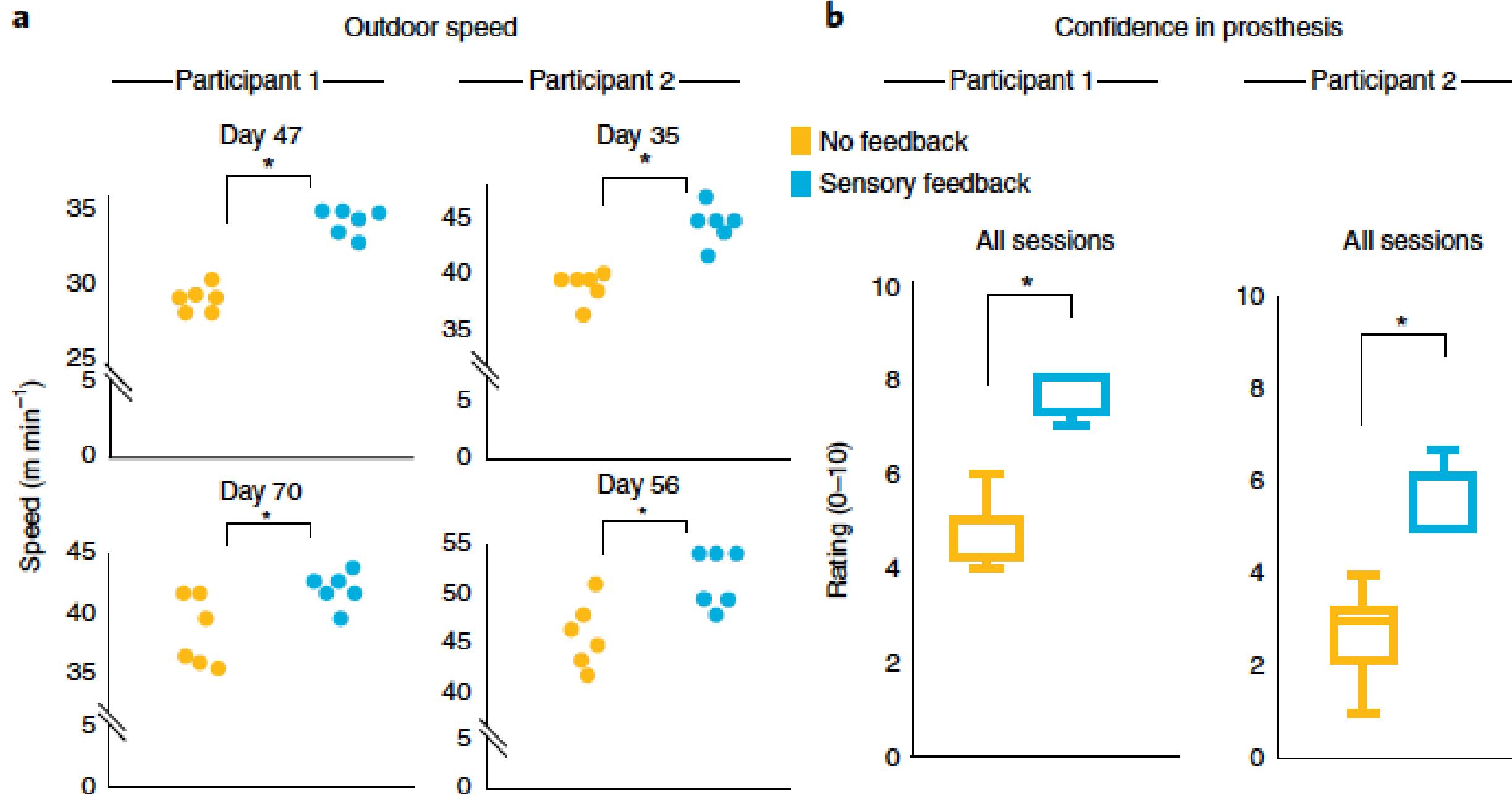


Petrini et al.,
Science Trans Med, 2019



Bidirectional neurocontrolled leg prostheses

Sensory feedback

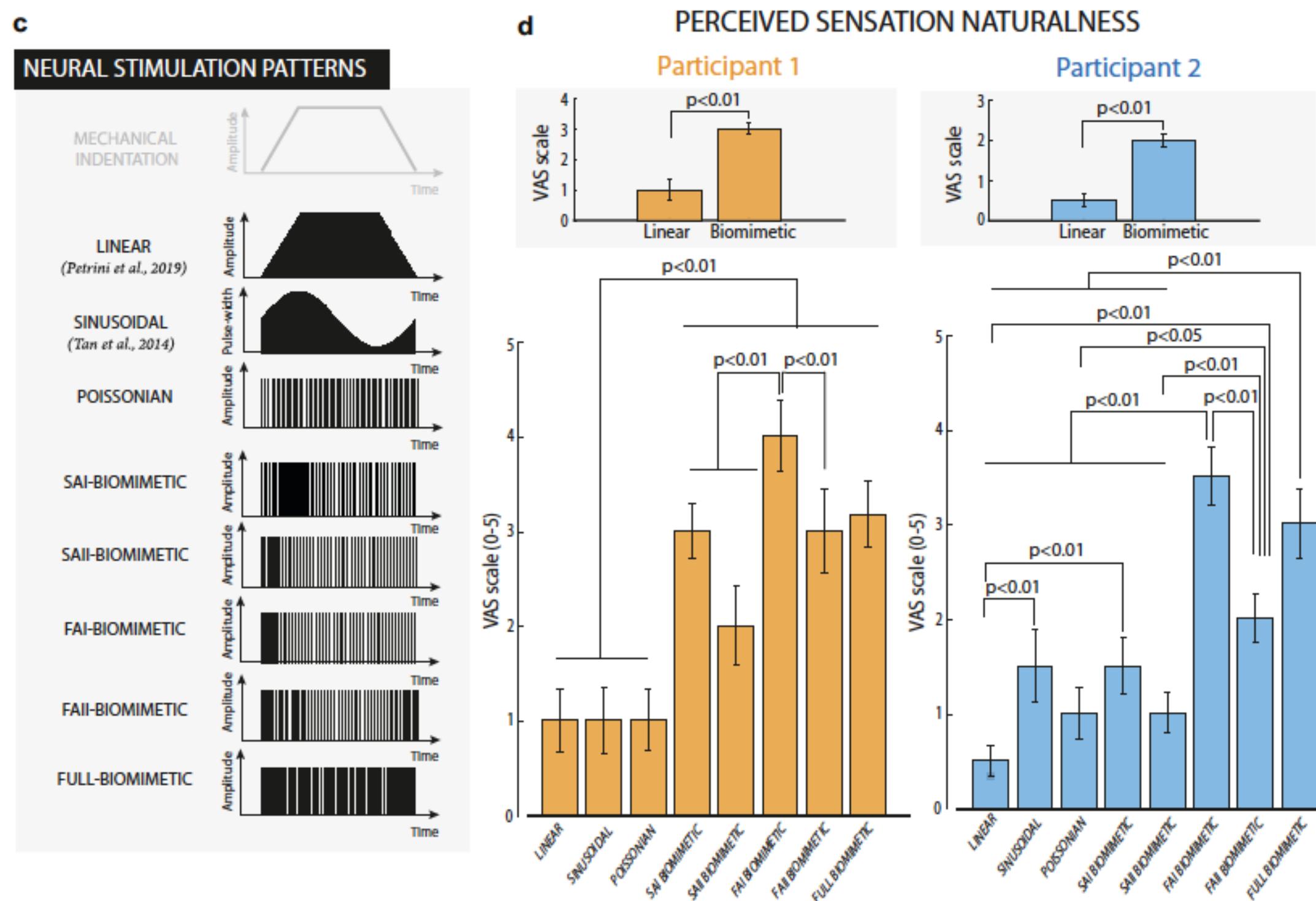
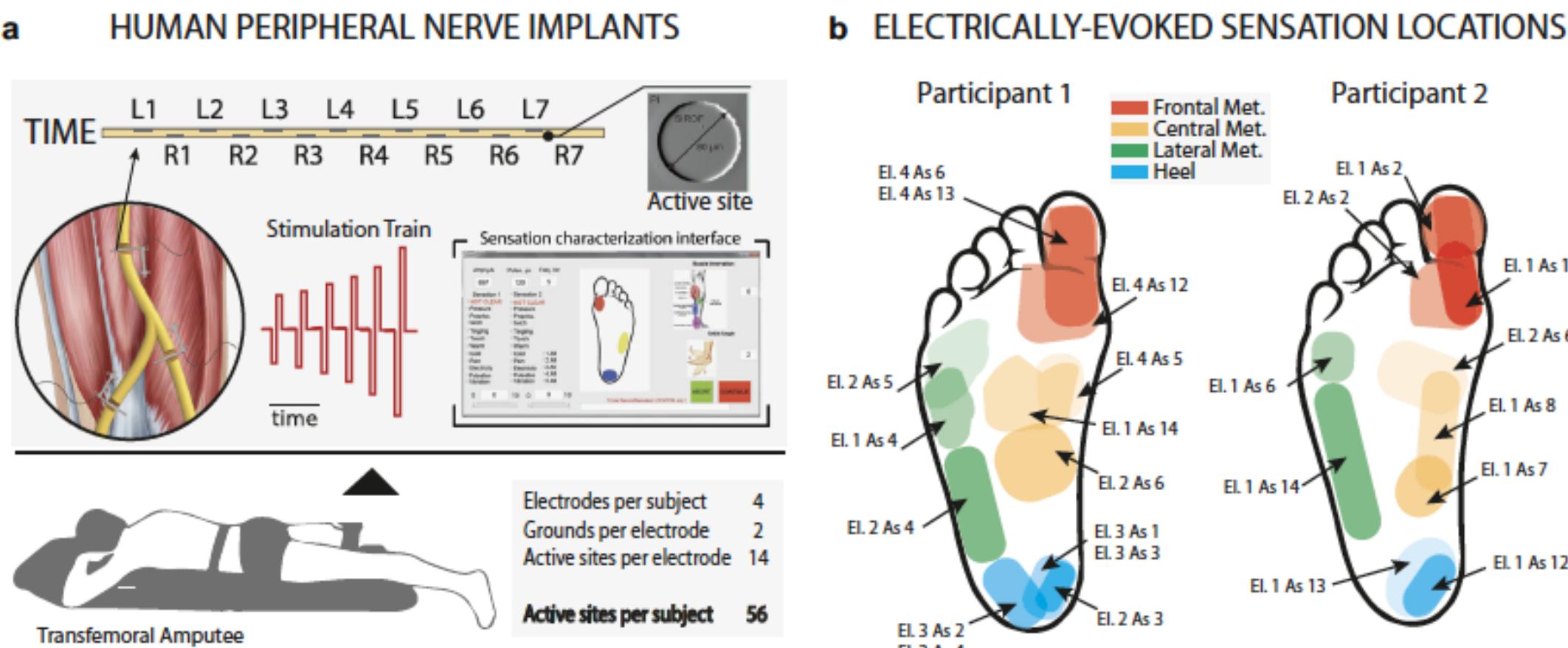


Walking speed and self-reported confidence increased while mental and physical fatigue decreased for both participants

Participants exhibited reduced phantom limb pain with neural sensory feedback.

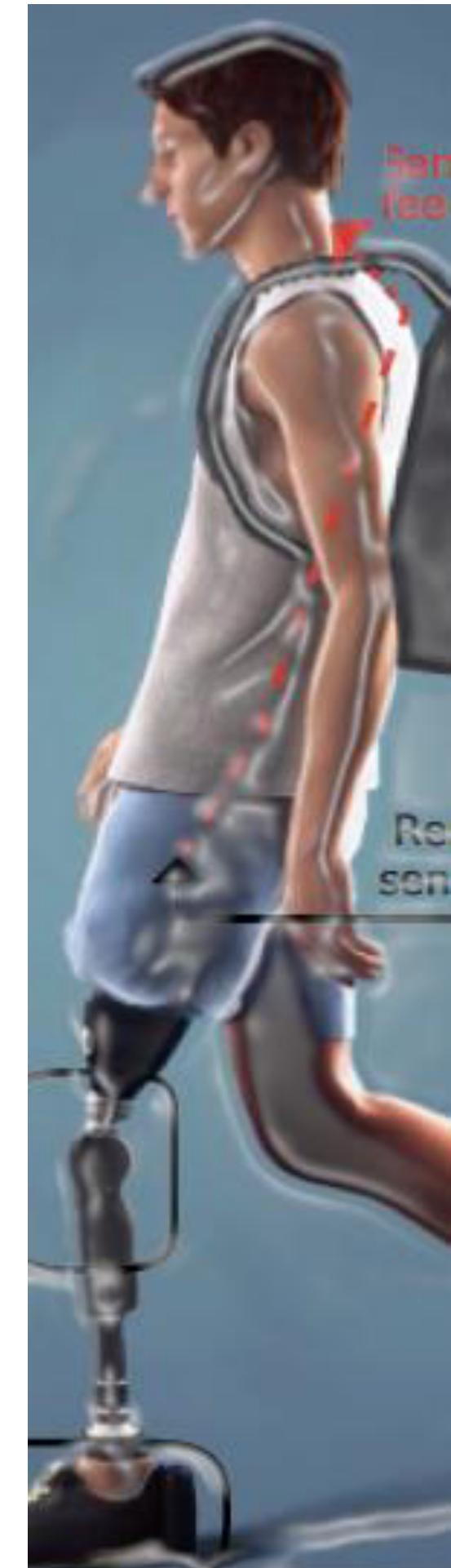
Bidirectional neurocontrolled leg prostheses

Sensory feedback



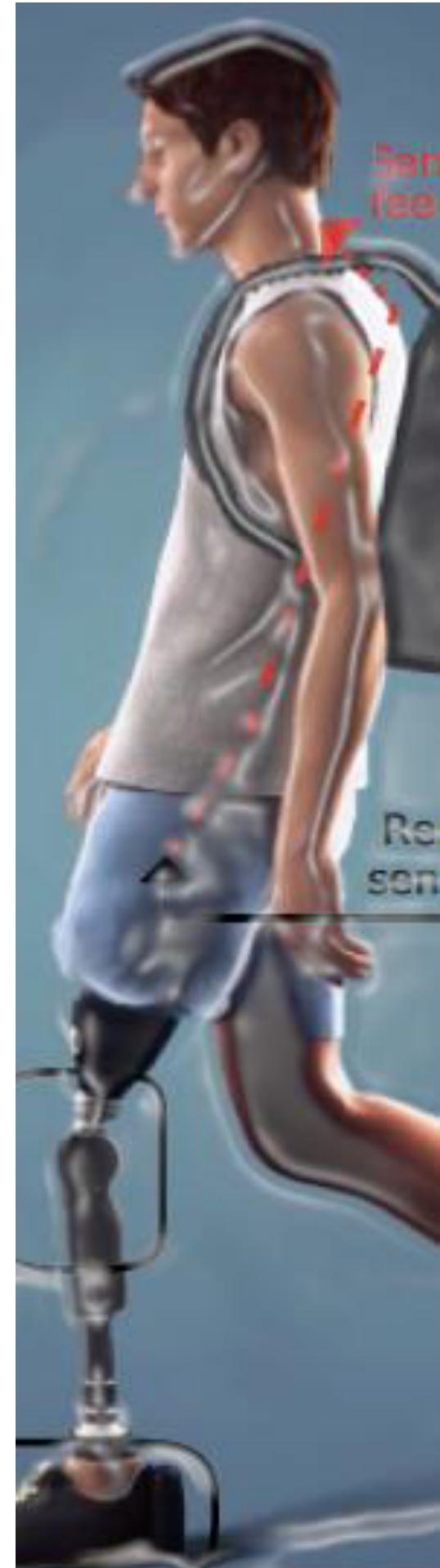
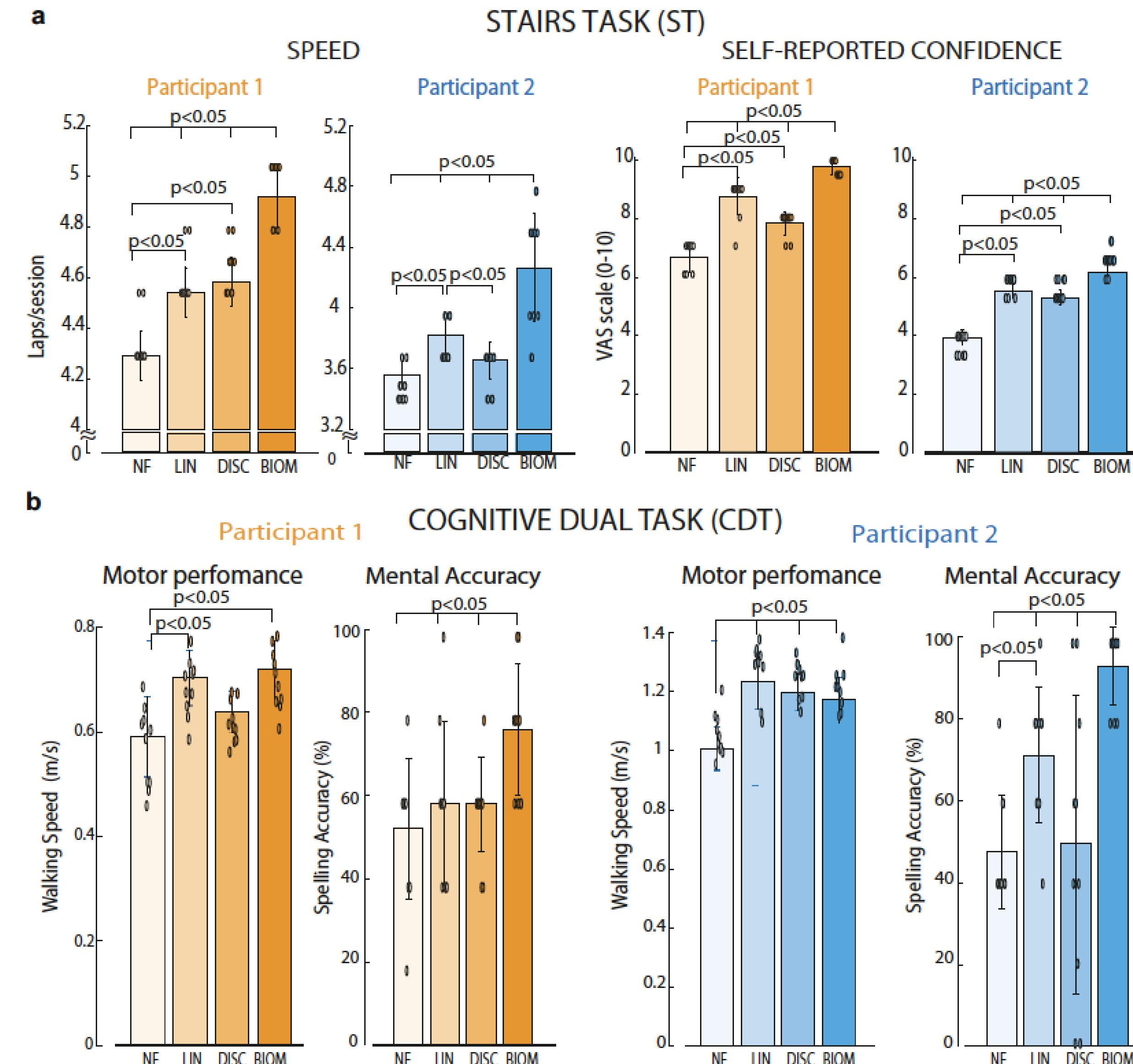
...This was probably caused by the different composition of the fibers activated by the electrode channels in the different foot regions.

...We hypothesized that not only the proportion of SA and FA fibers is relevant but also their role in encoding touch information in that specific region.



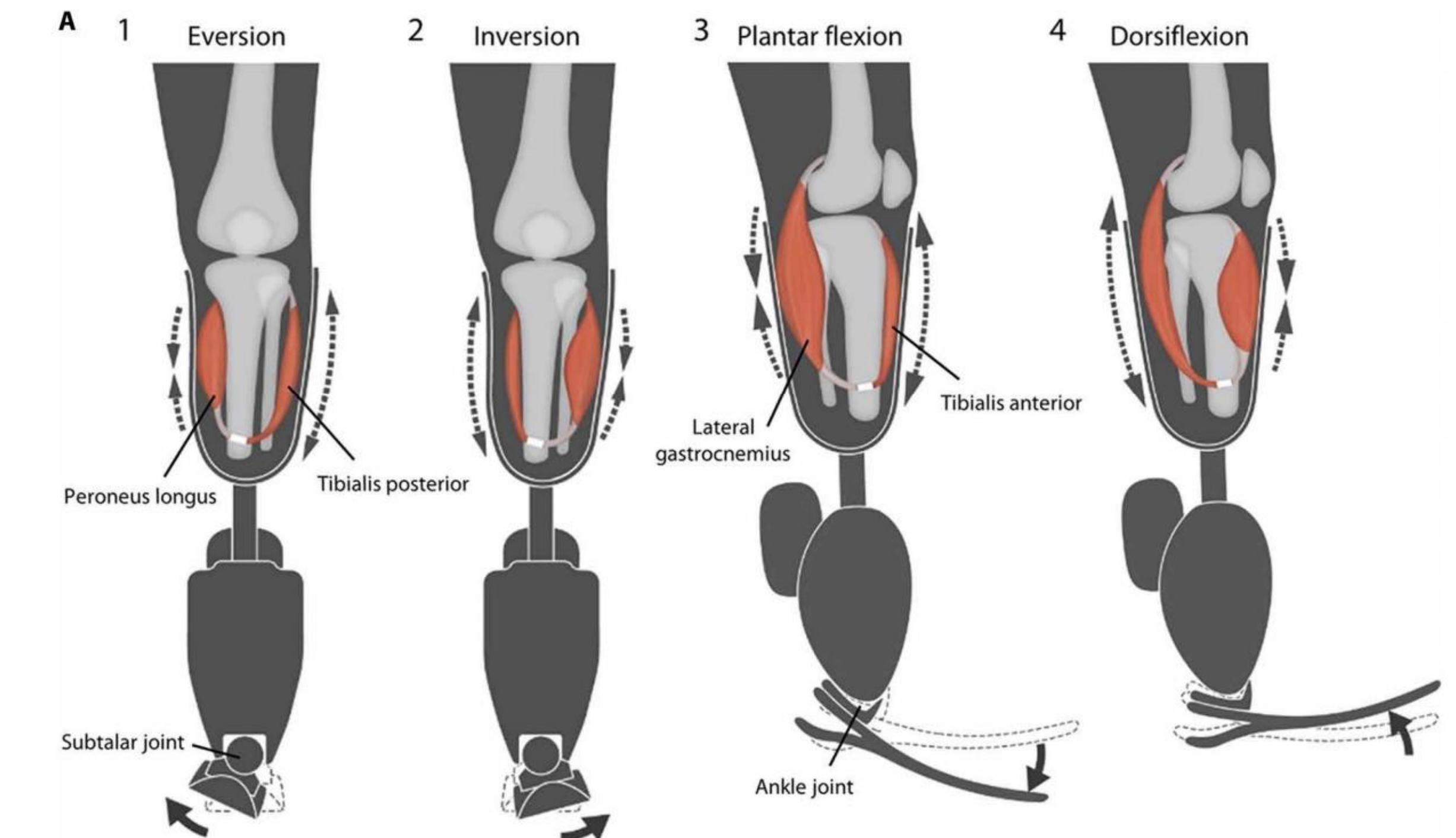
Bidirectional neurocontrolled leg prostheses

Sensory feedback

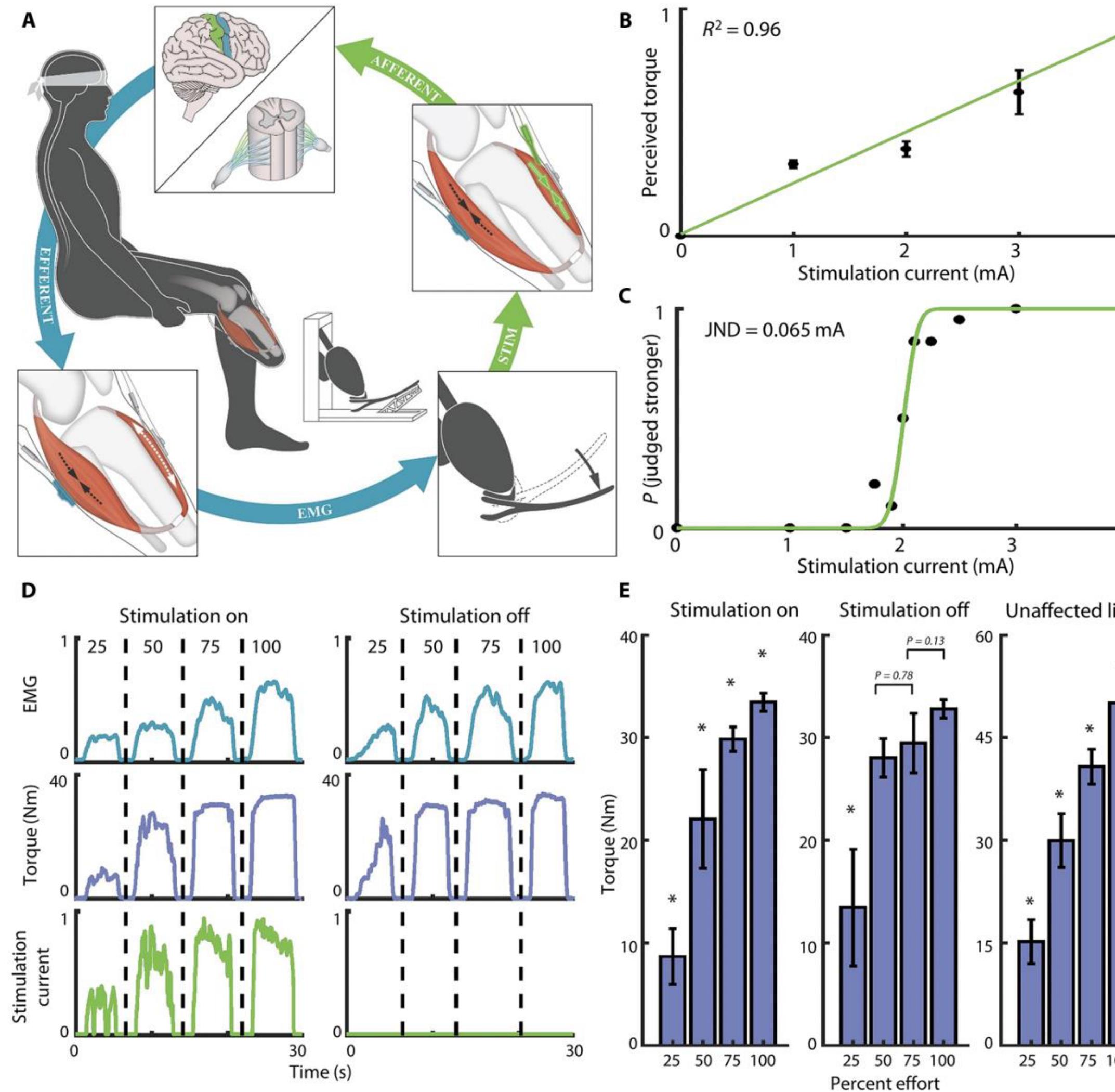


Agonist-antagonist myoneural interface

- As a methodology of improving efferent (neural pathways that relay commands from the central nervous system to a muscle or other end organ) prosthetic control and providing afferent proprioceptive sensation, we present an agonist-antagonist myoneural interface (AMI)
- An AMI is made up of an agonist and an antagonist muscle tendon connected mechanically in series: When the agonist contracts, the antagonist is stretched and vice versa
- The purpose of an AMI is to control and interpret proprioceptive feedback from a bionic joint.

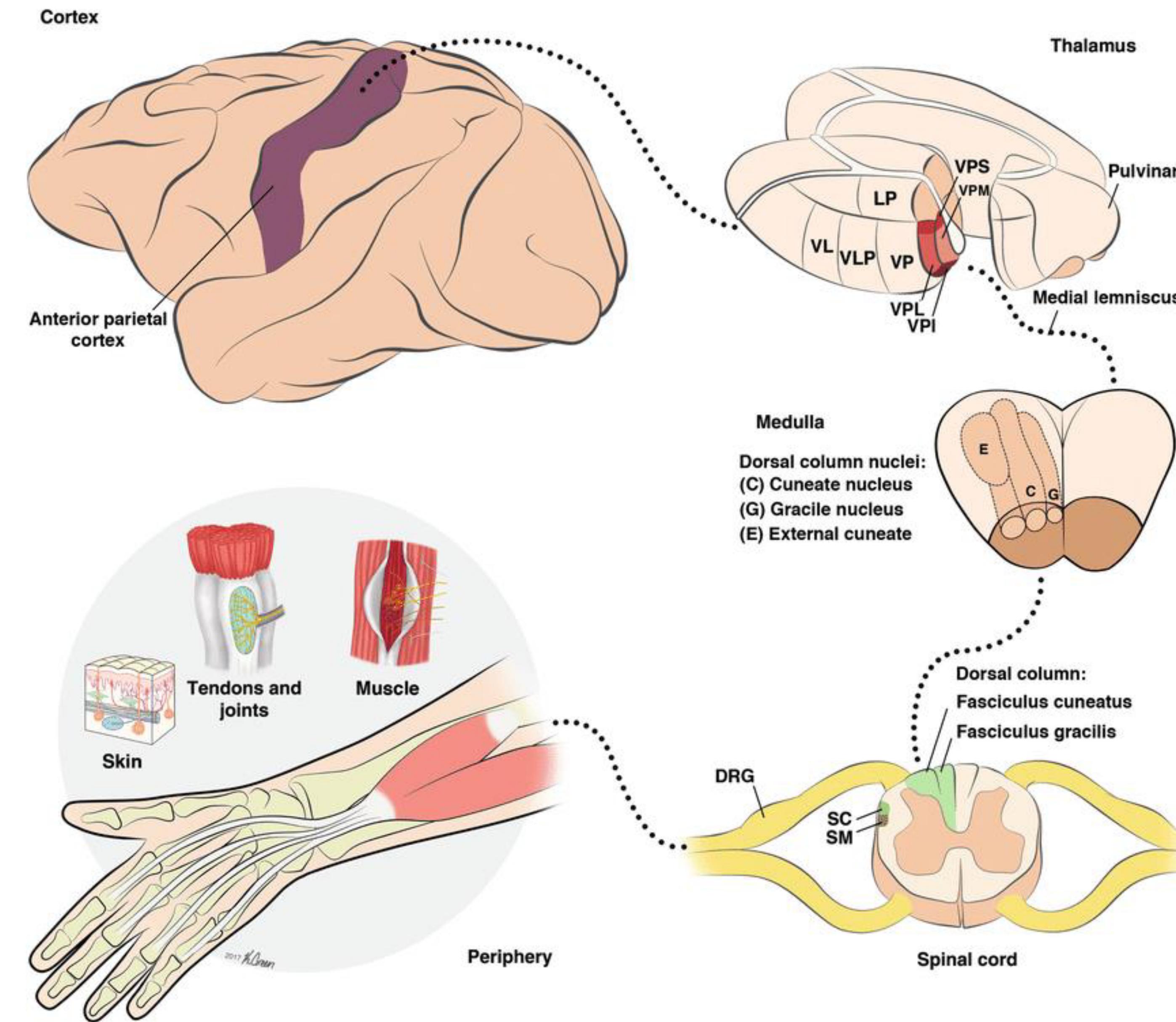


Agonist-antagonist myoneural interface – Closed-loop torque control

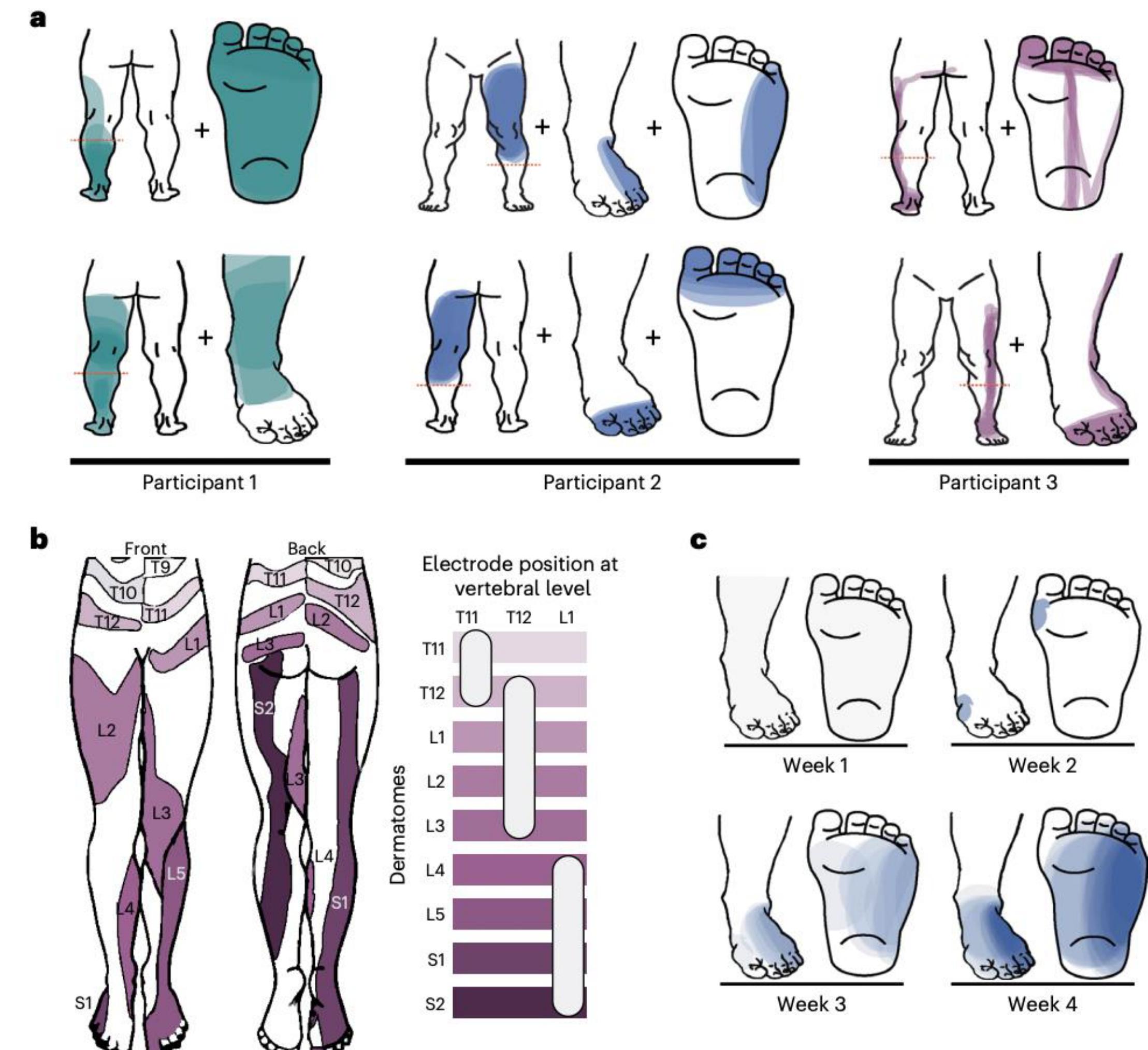


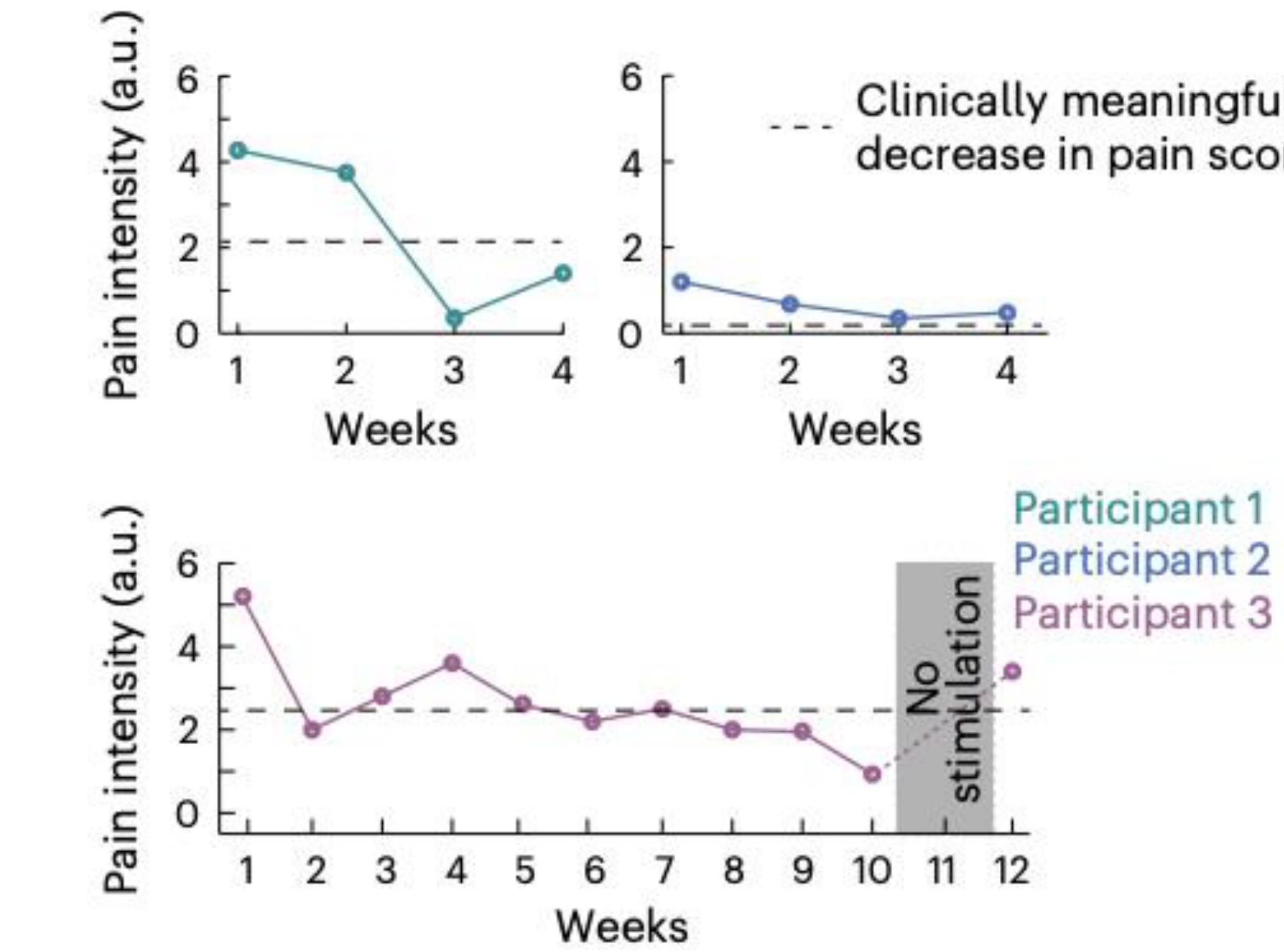
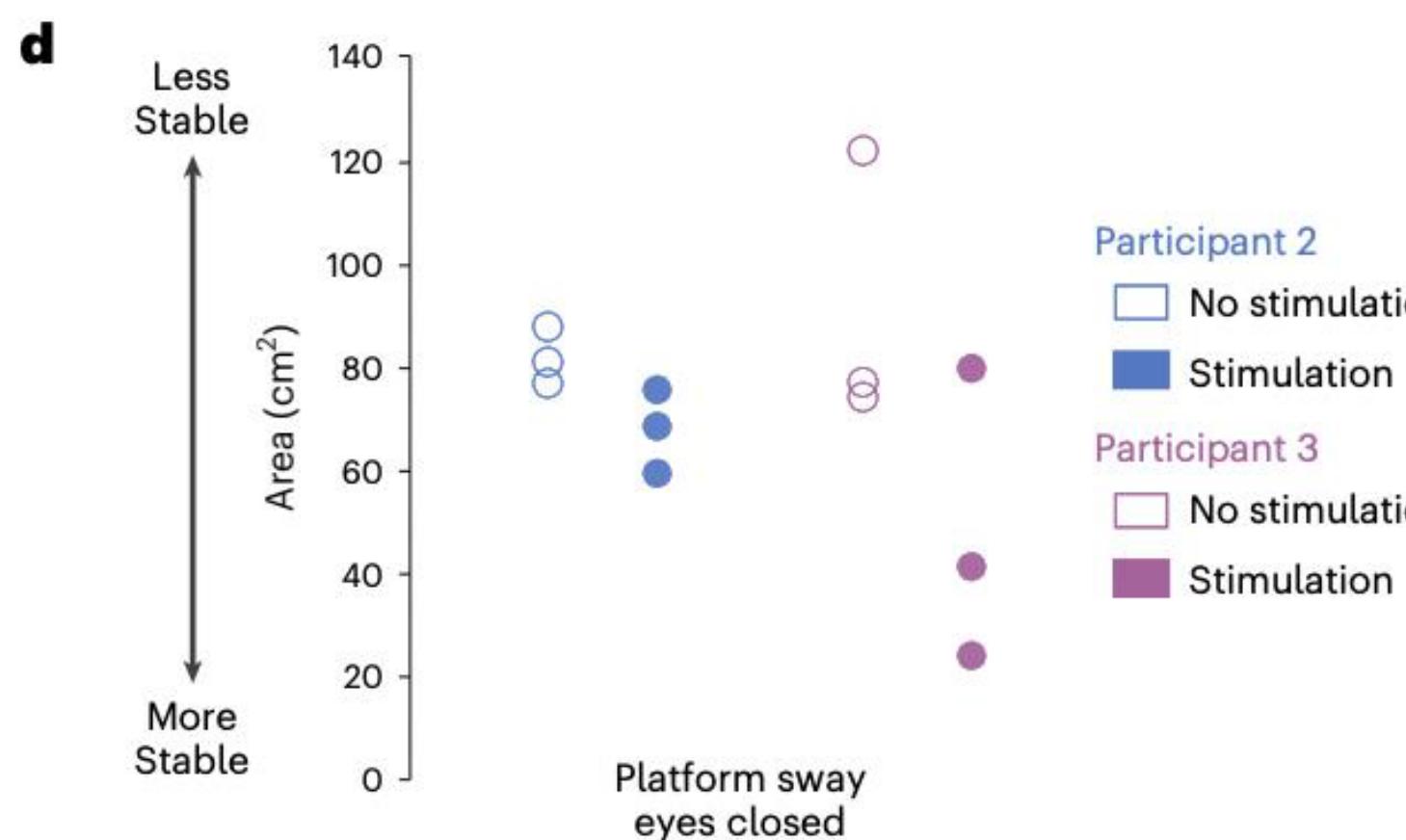
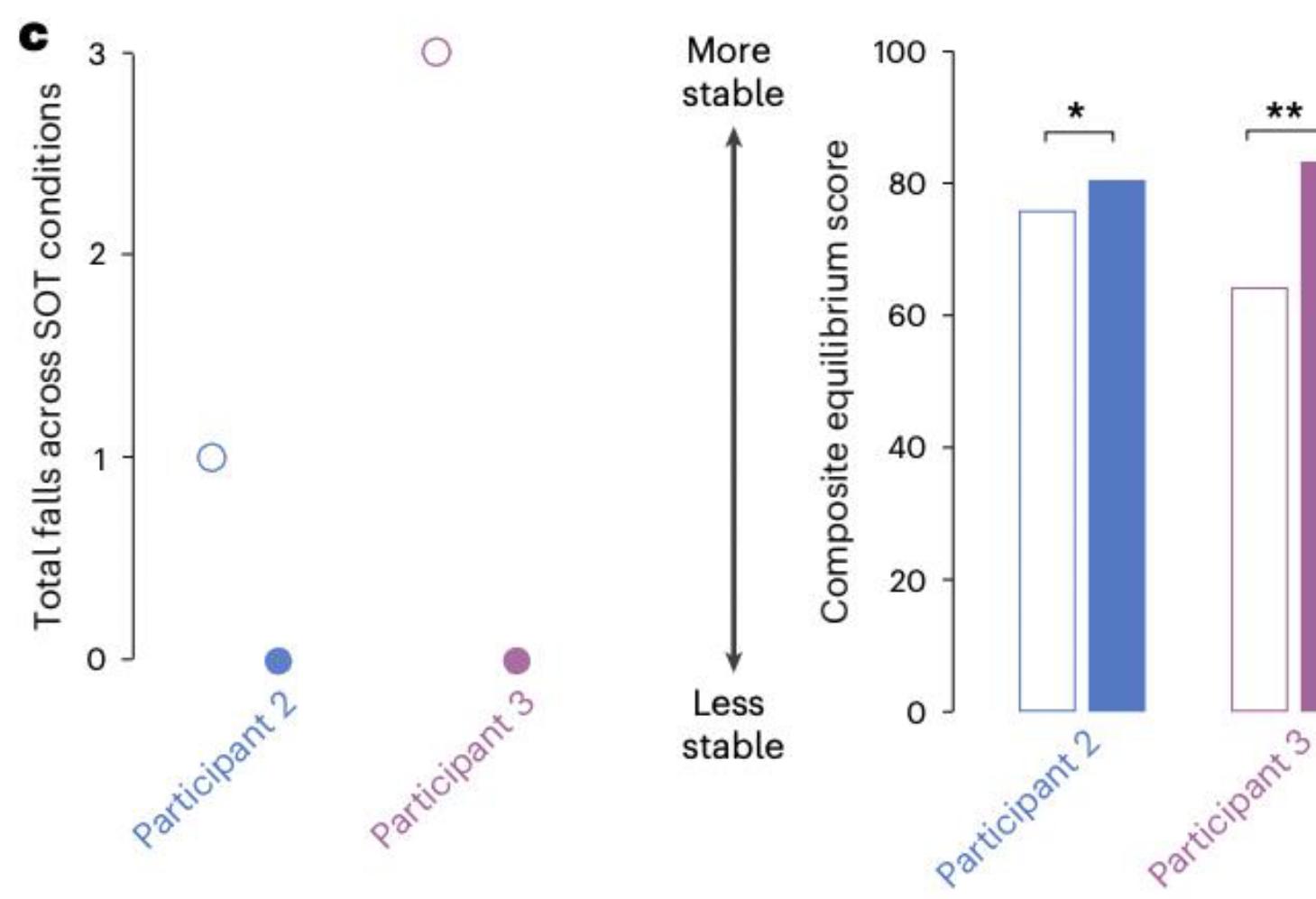
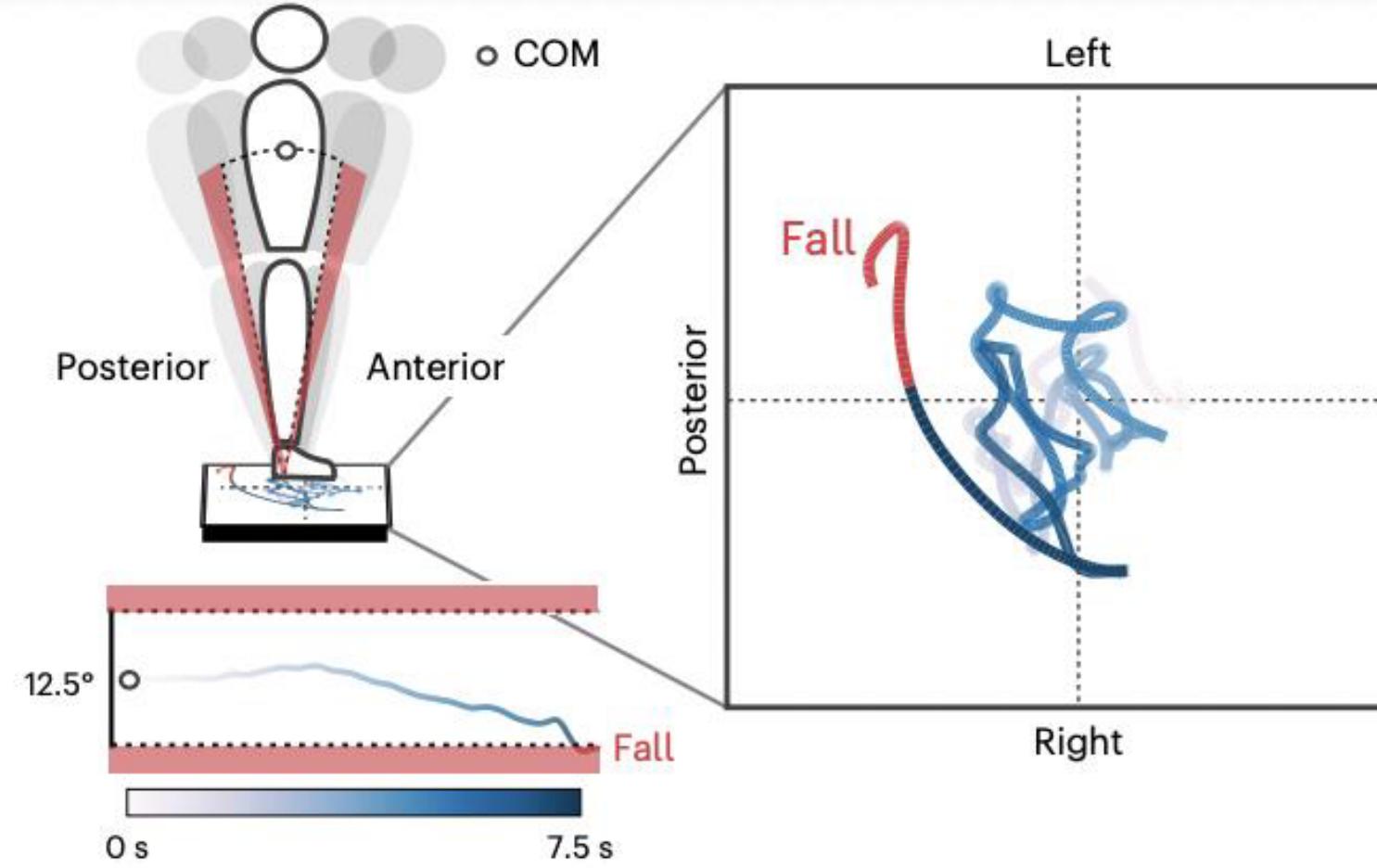
- (A) Schematic of the prosthesis-in-the-loop control architecture, in which afferent feedback of prosthetic joint torque is provided via FES of the antagonist muscle. The patient perceives this stimulation as a natural sensation of ankle torque
- (B) Magnitude estimation of perceived dorsiflexion torque as a function of stimulation current delivered to the tibialis anterior
- (C) Discrimination performance as a function of differences in stimulation current
- (D) Representative sample traces of lateral gastrocnemius EMG (blue), torque (purple), and stimulation current (green) during closed-loop torque control trials for the “stimulation on” ($n = 79$ total trials) and “stimulation off” ($n = 79$ total trials) cases
- (E) Summary data for closed-loop torque control trials in each of the stimulation on ($n = 79$ trials), stimulation off ($n = 79$ trials), and “unaffected limb” ($n = 80$ trials) cases

Human touch system



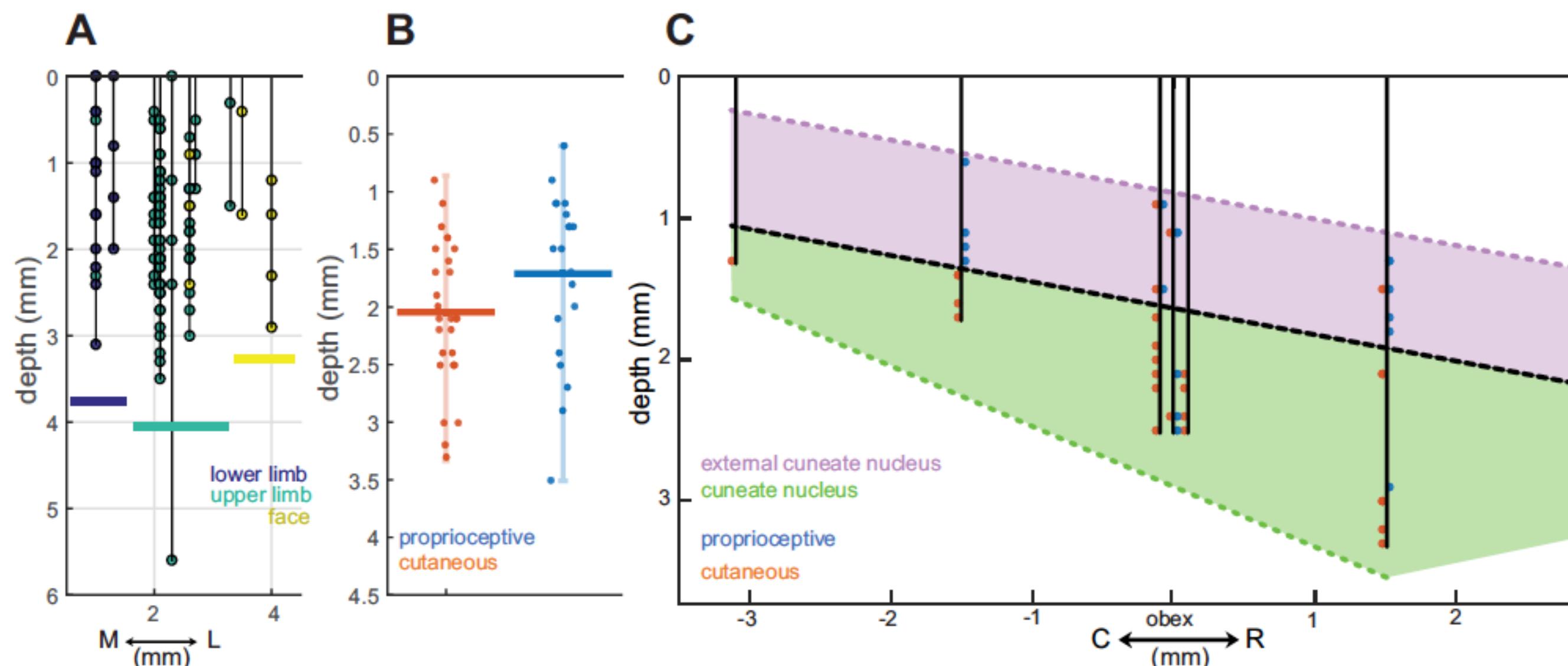
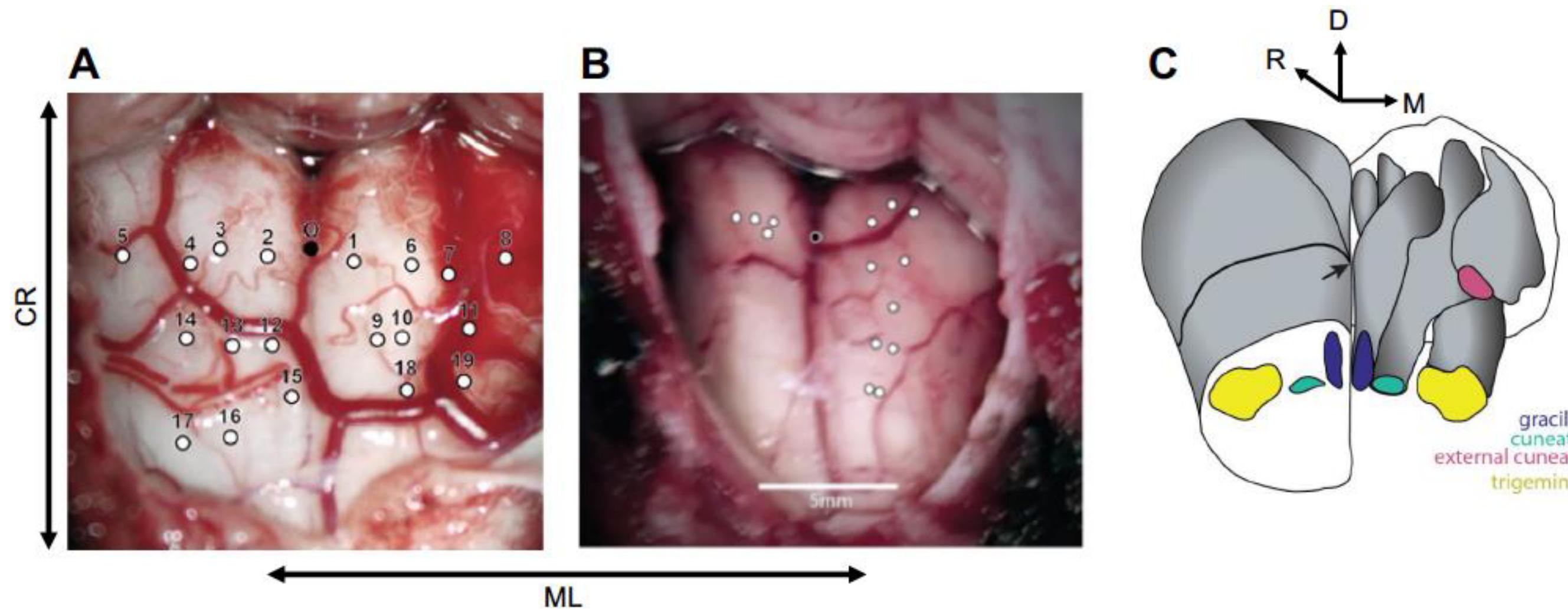
Spinal cord stimulation



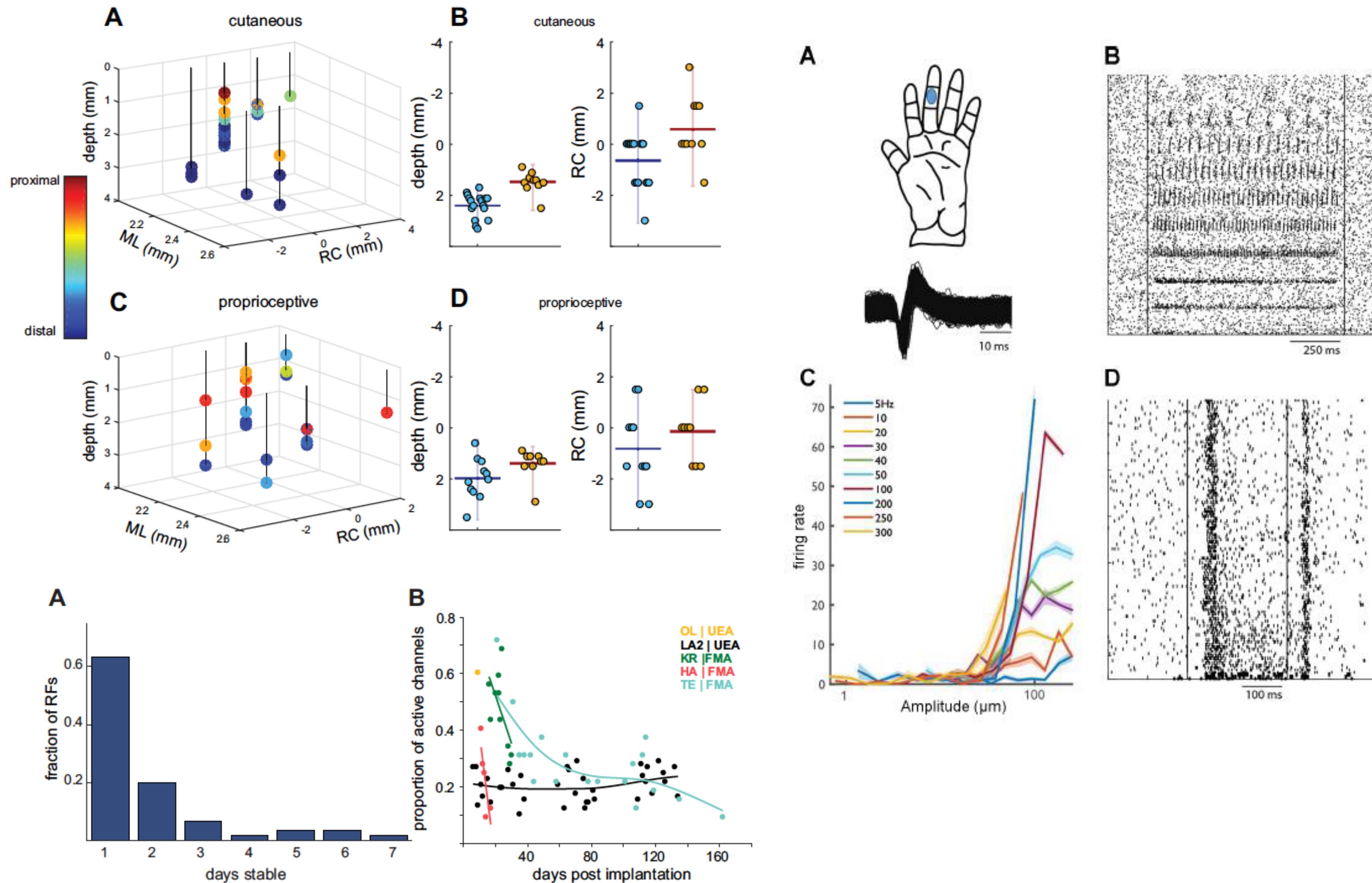


Performance good (maybe a bit less than PNS stimulation) BUT with a clinically accepted surgical procedure

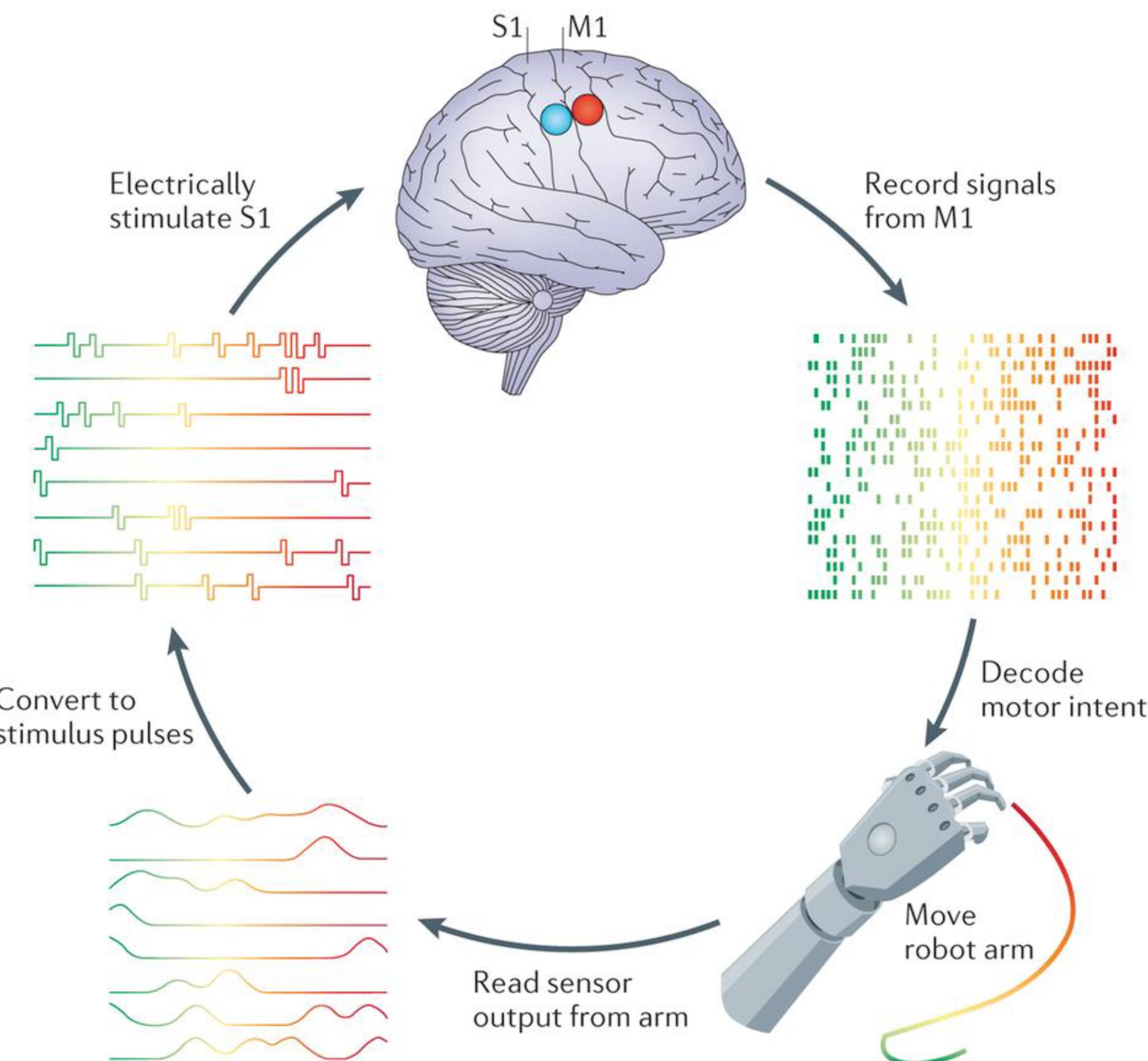
Stimulation of the cuneate for sensory feedback



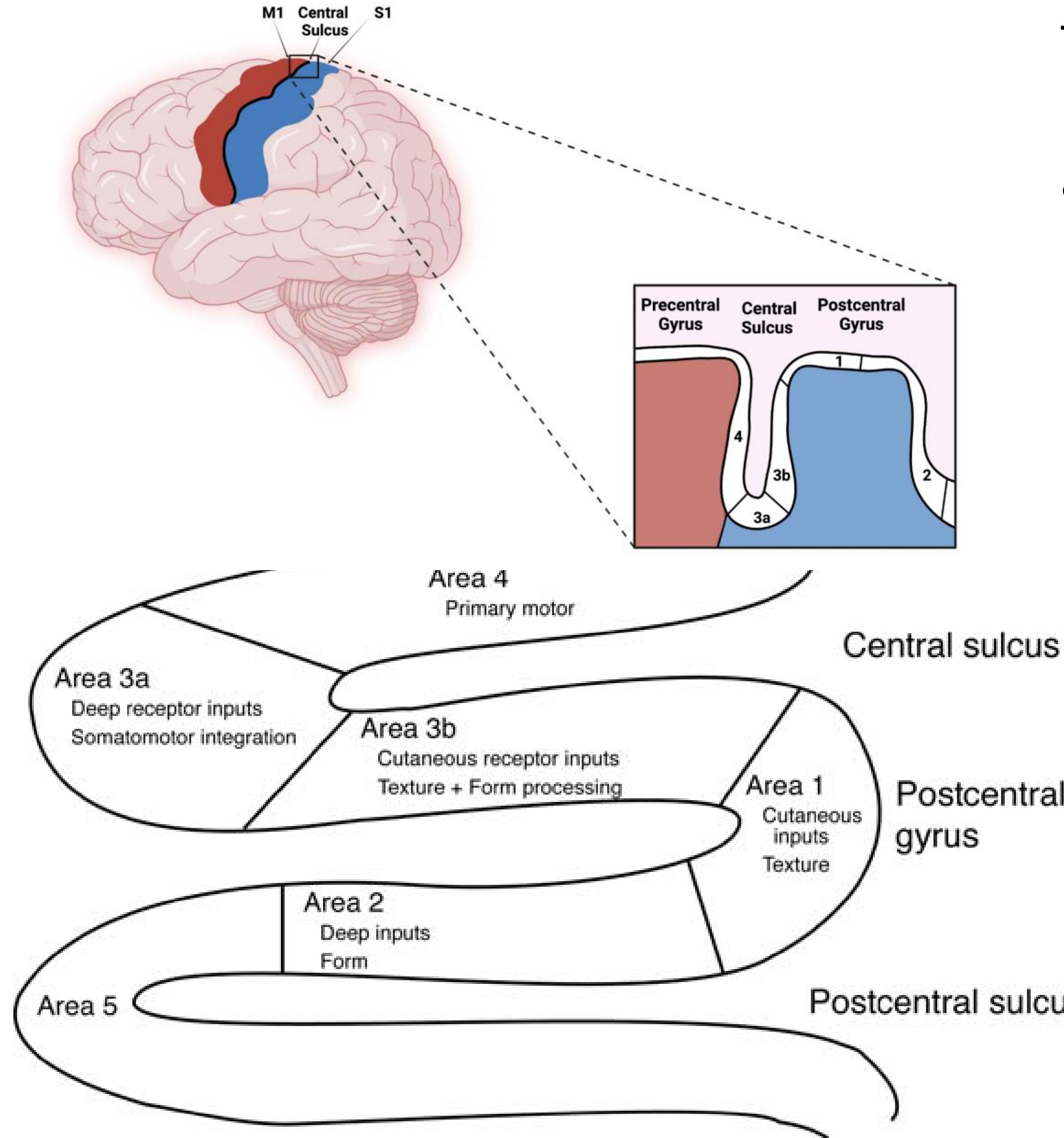
Stimulation of the cuneate for sensory feedback



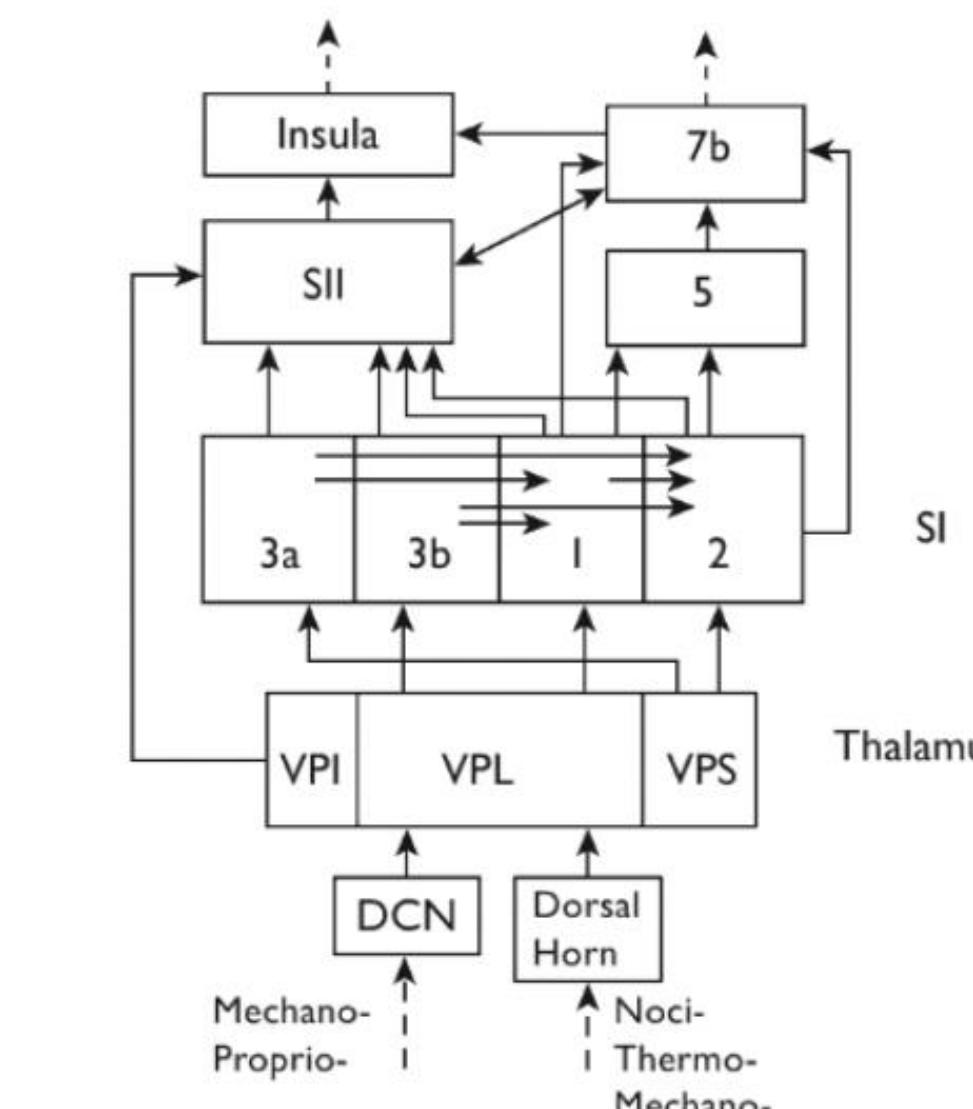
Brain-to-machine-to-brain interface



Brain-to-machine-to-brain interface

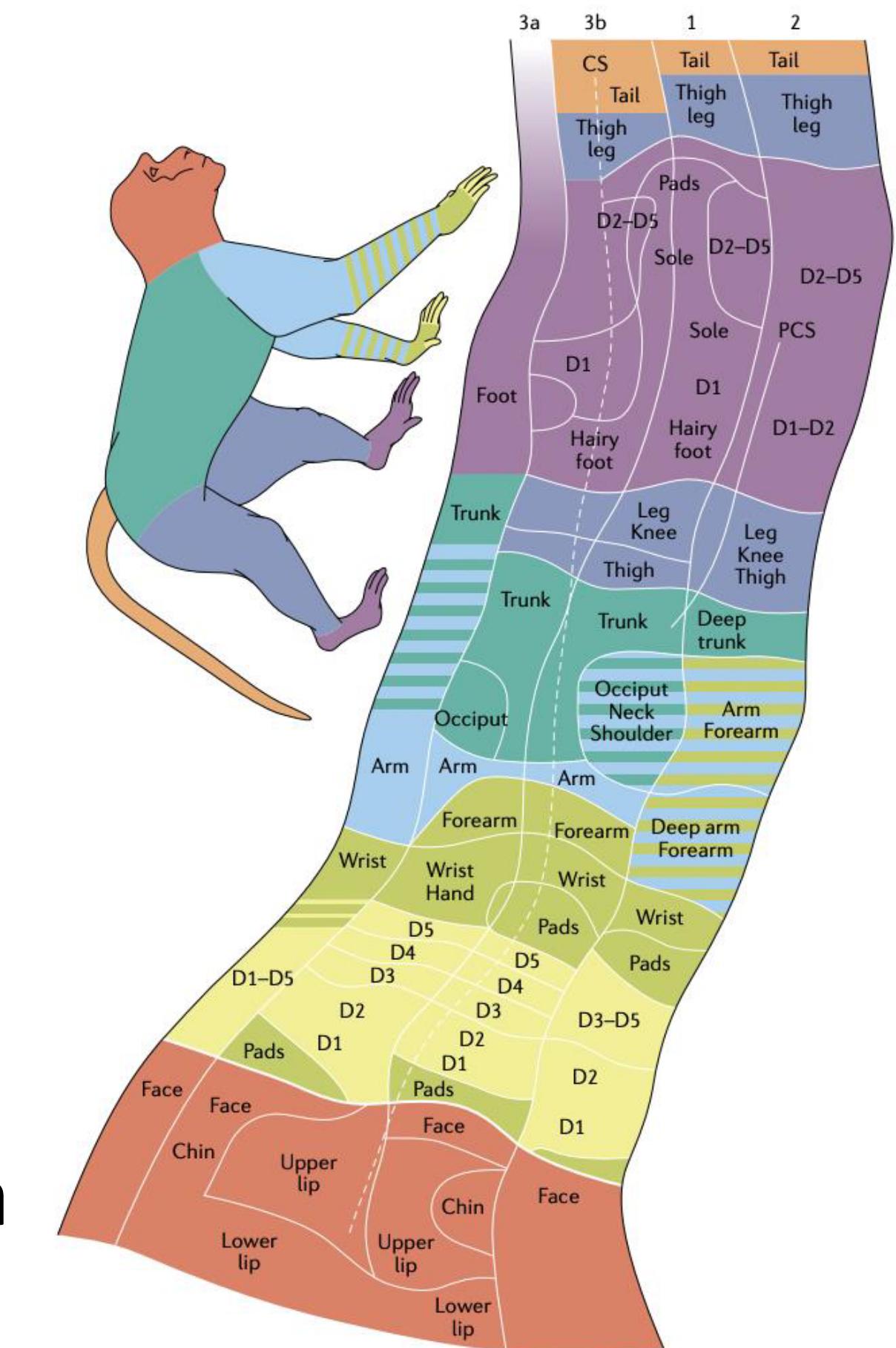


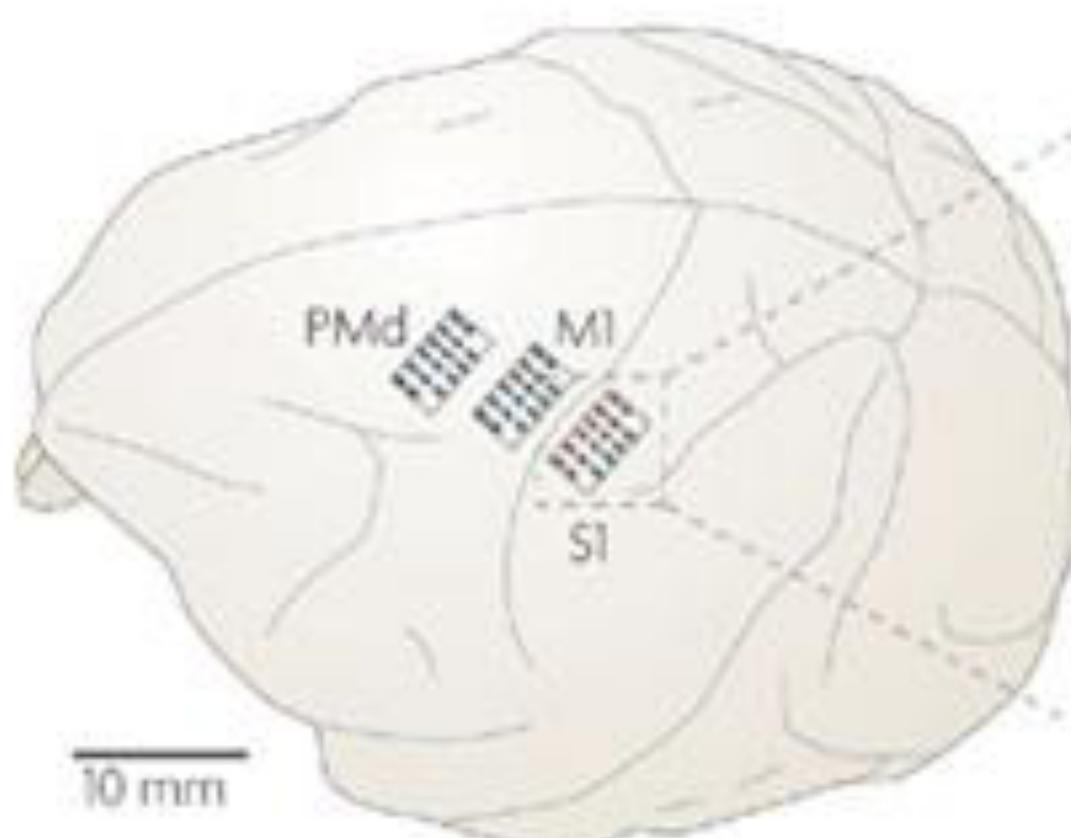
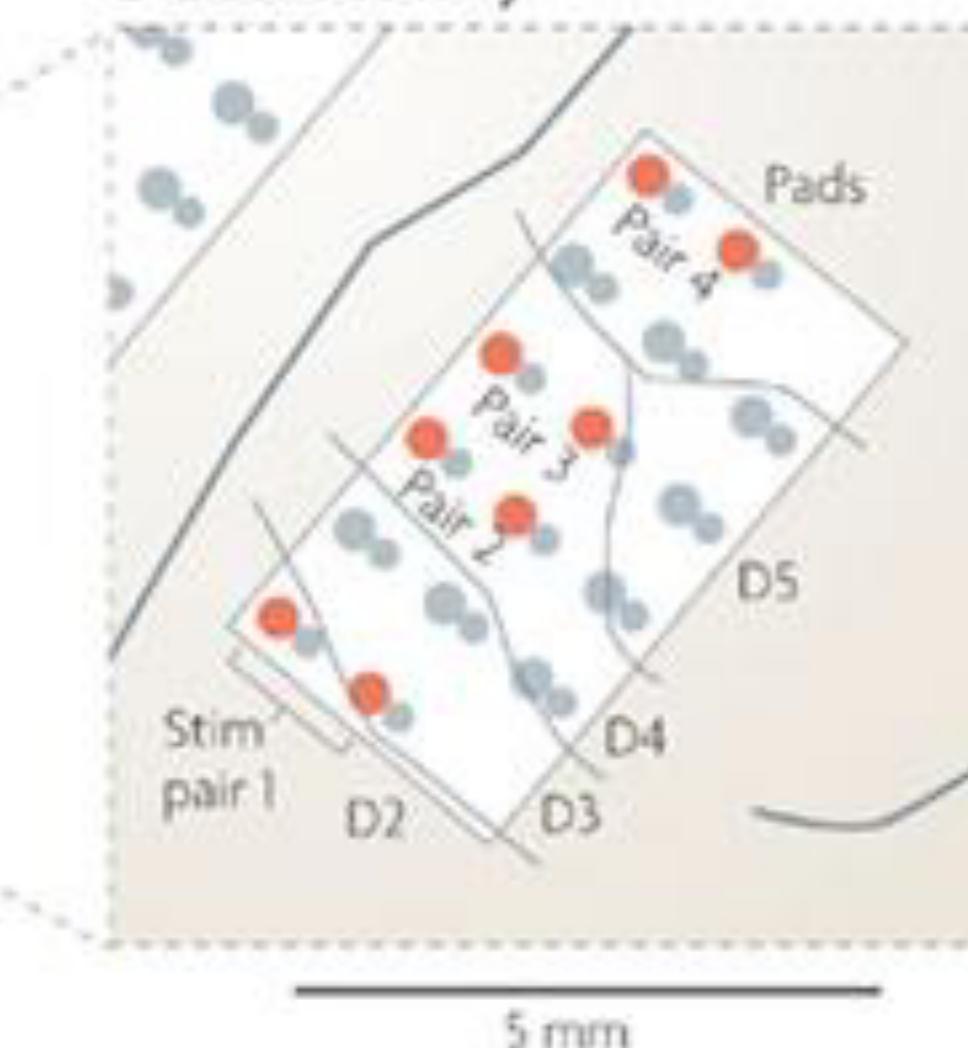
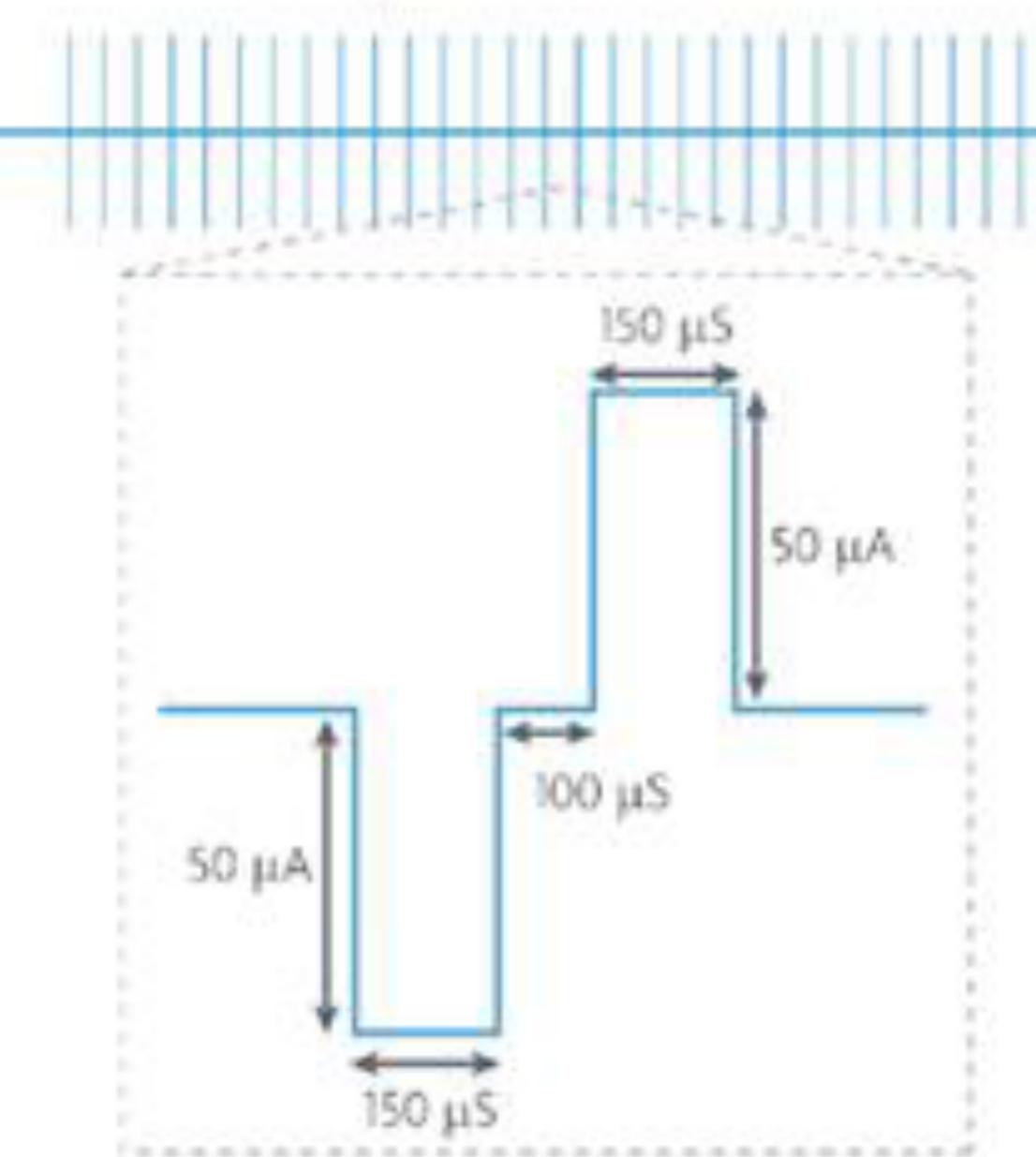
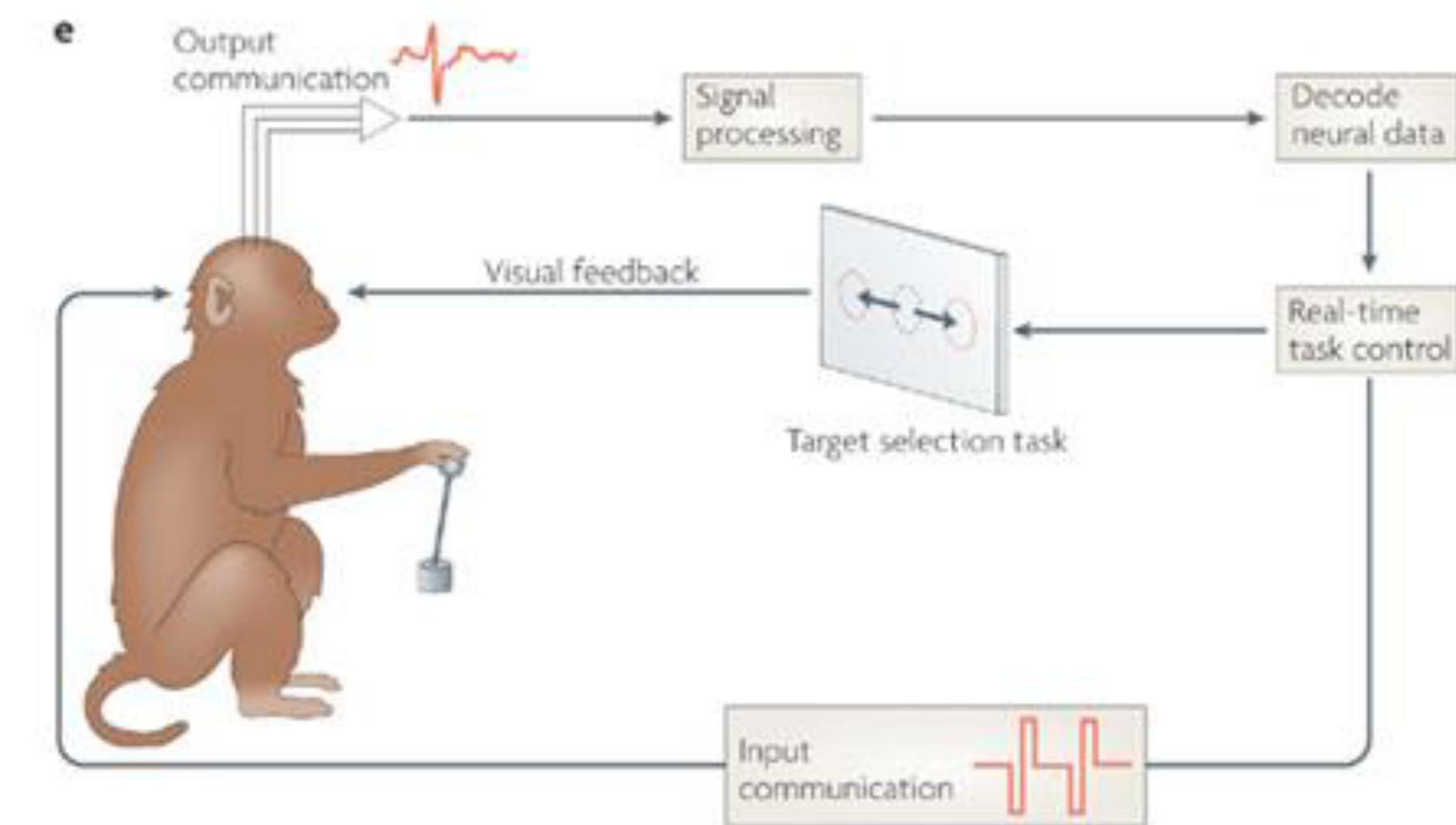
The somatosensory homunculus comprises four body maps, one each in Brodmann areas 3a, 3b, 1 and 2, which exhibit proprioceptive responses (area 3a), cutaneous responses (areas 3b and 1) or both (area 2).



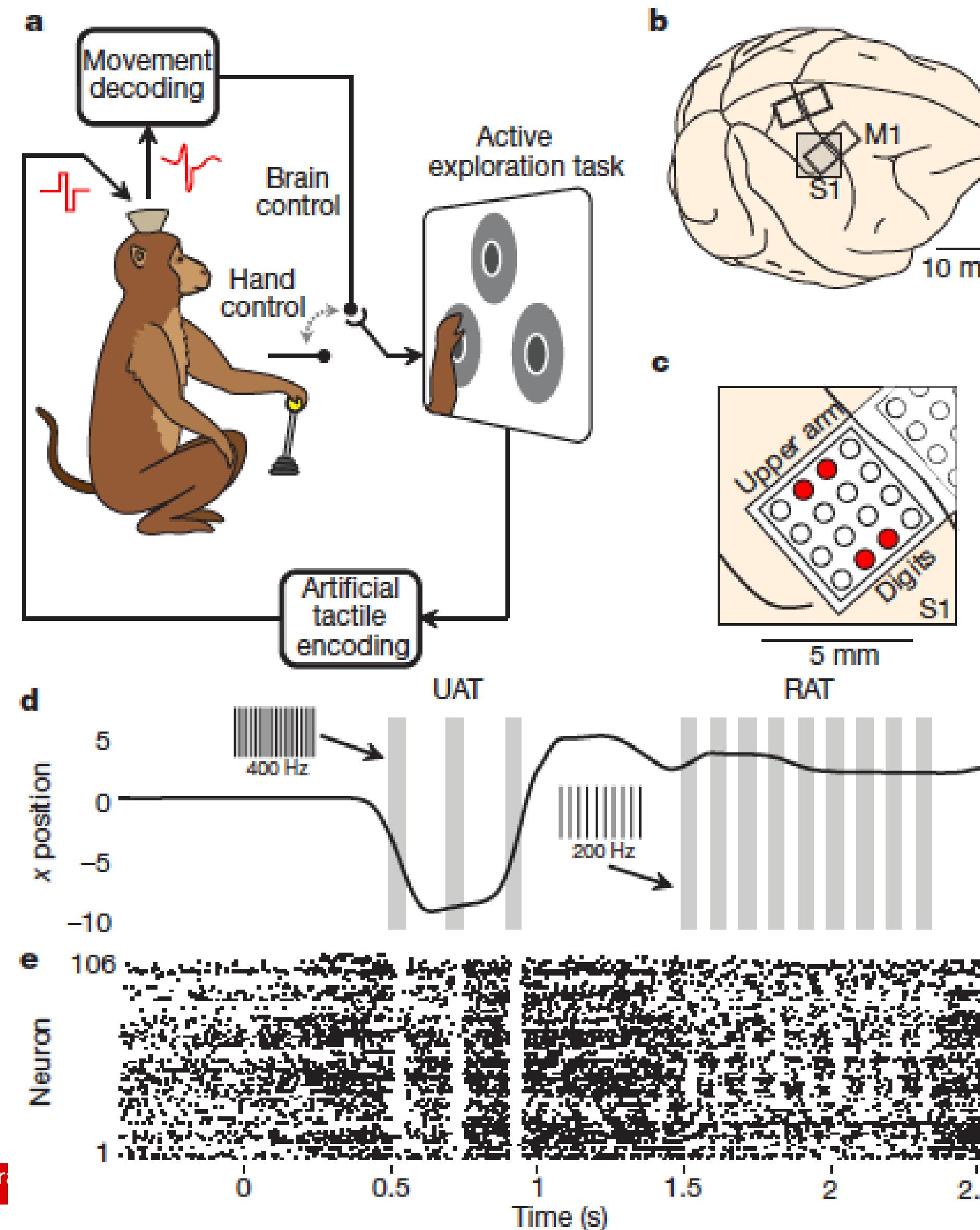
Sobinov and Bensmaia

Nat Rev Neuro, 2021



a Implantation sites**b Electrode array****c Receptive fields****d Stimulation pattern****e**

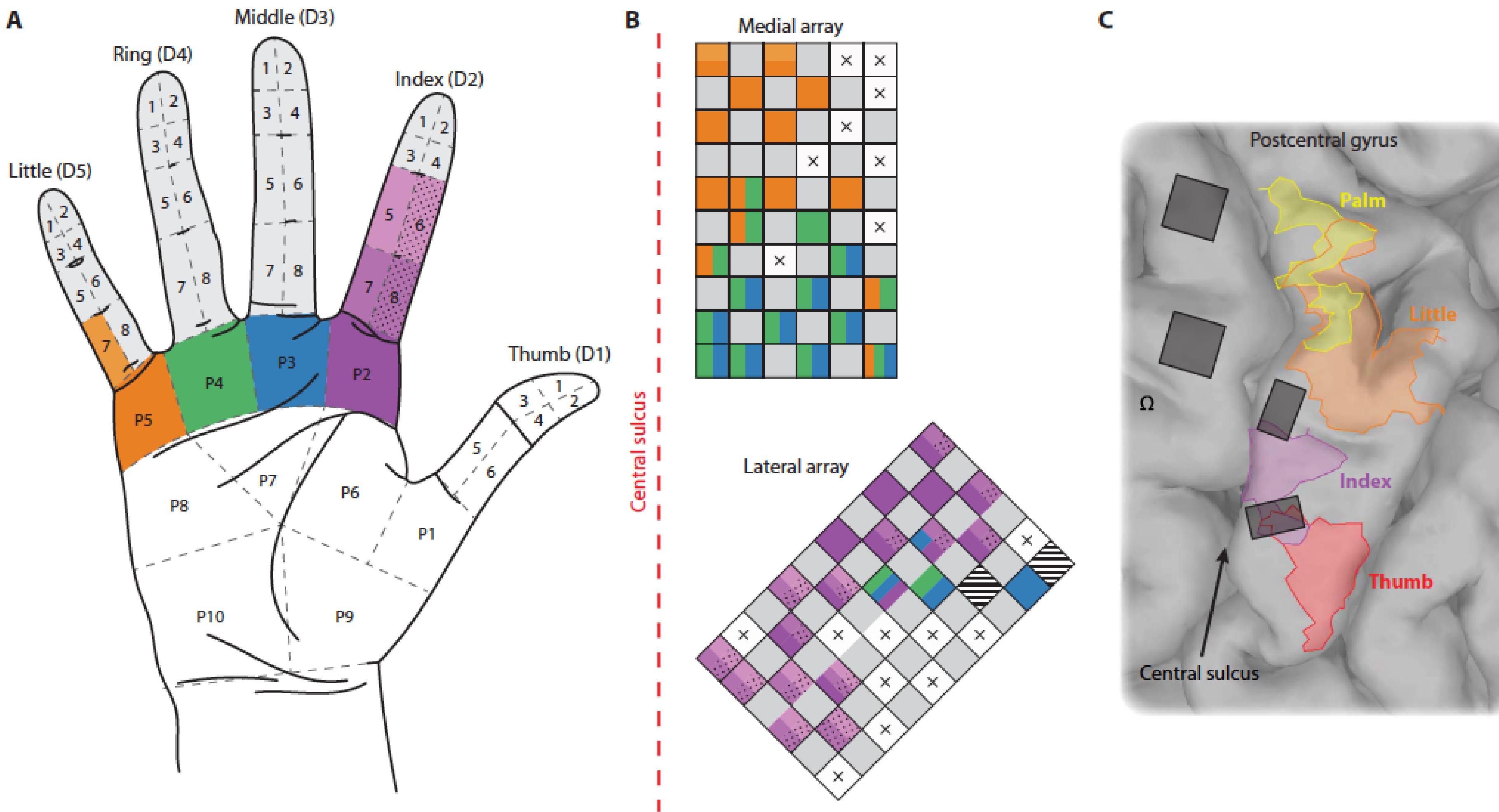
Intracortical sensory feedback



Intracortical sensory feedback is possible but the performance are still limited

O'Doherty et al., 2011

Brain-to-machine-to-brain interface in a quadriplegic subject



Brain-to-machine-to-brain interface in a quadriplegic subject

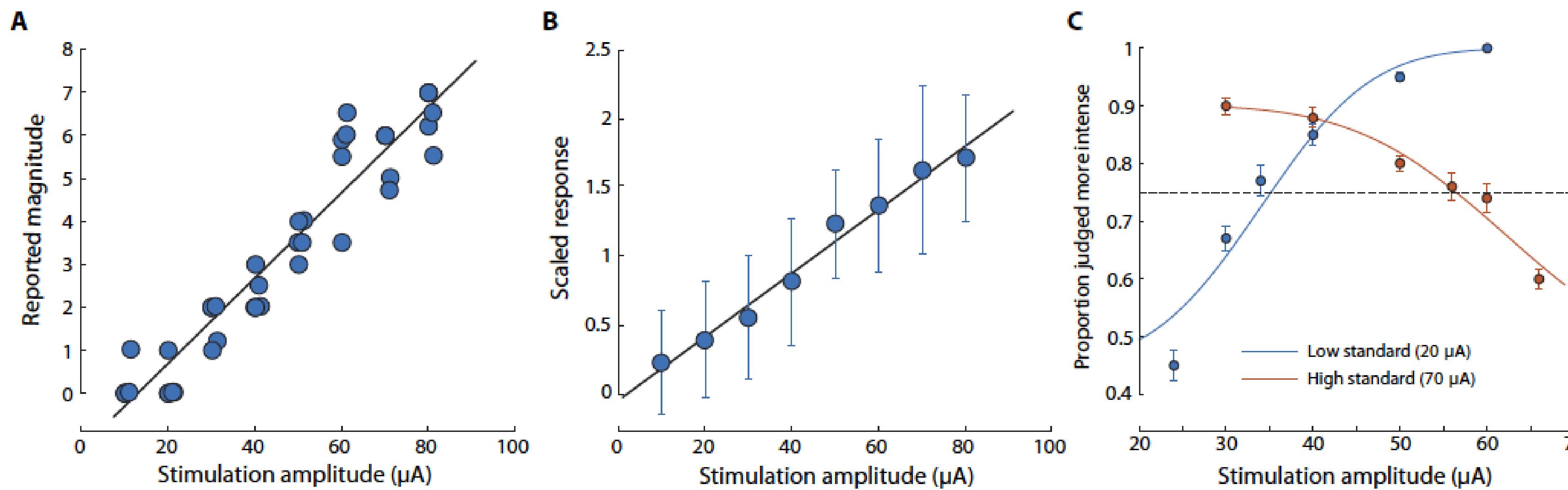
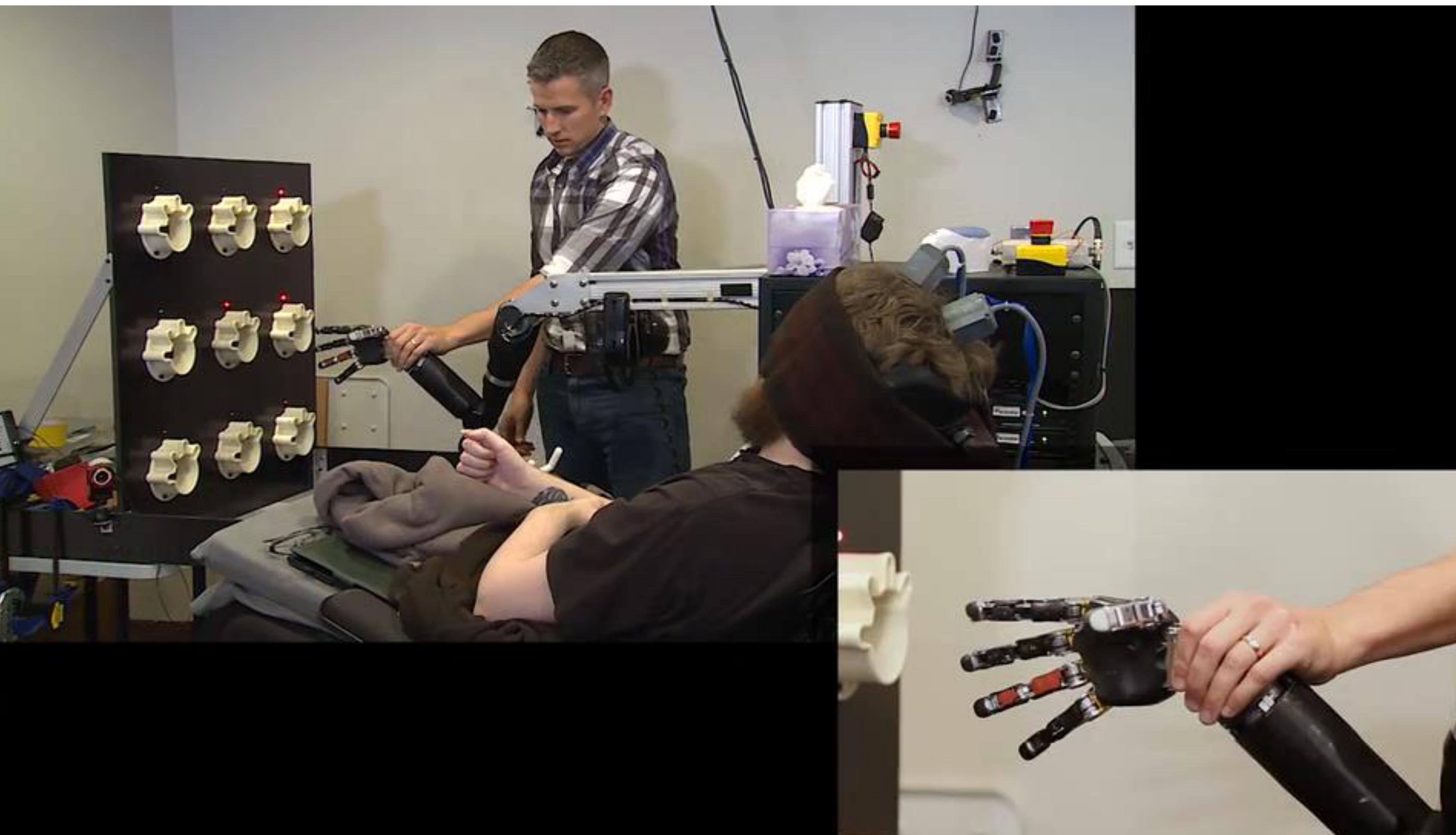


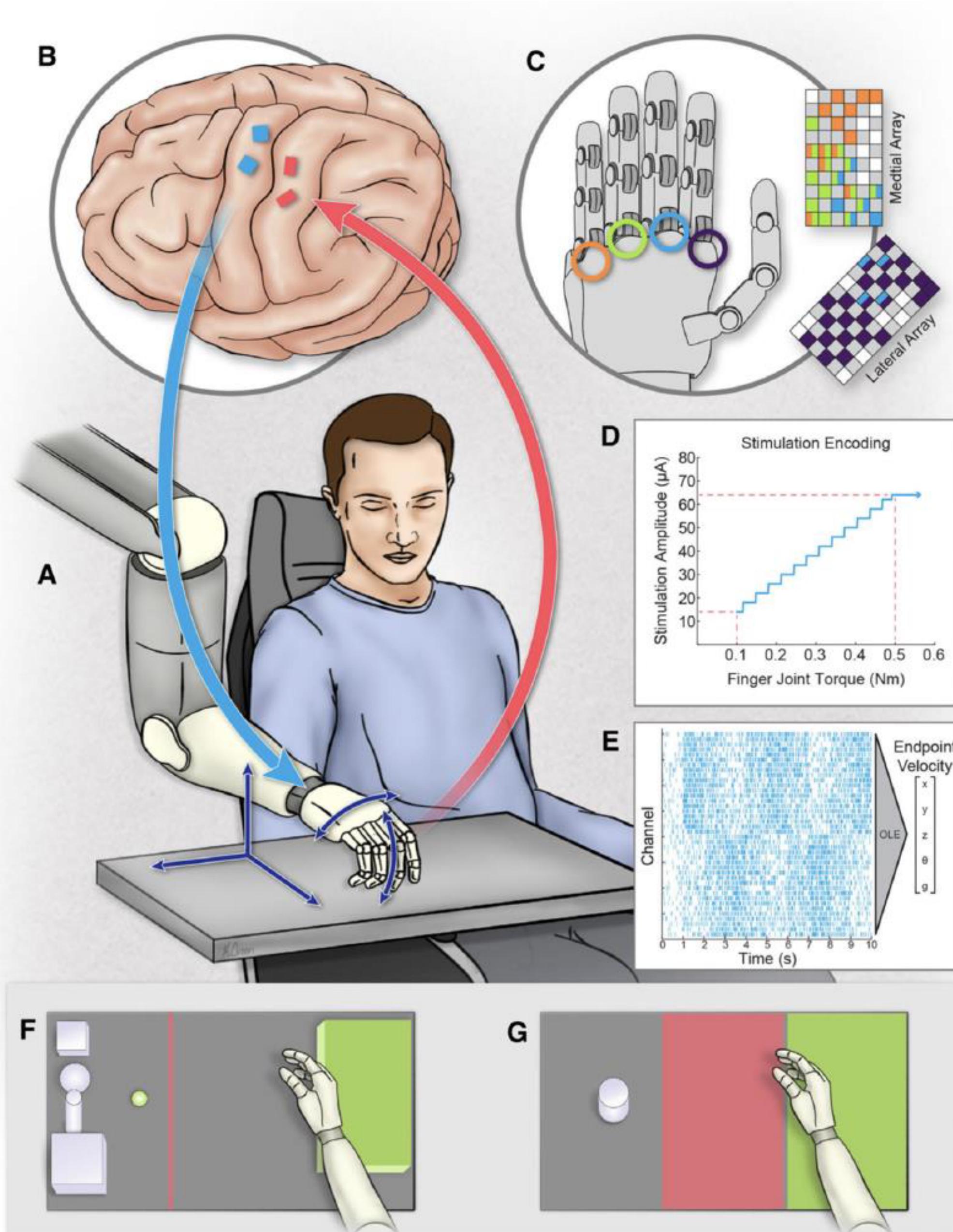
Table 2. Accuracy of prosthetic finger discrimination. The percentage of times that sensations were reported to originate from a specific finger (columns) when each prosthetic finger was touched (rows).

	Reported D2	Reported D3	Reported D4	Reported D5
Actual D2	$96.9 \pm 7.2\%$	$1.5 \pm 5.3\%$	$1.5 \pm 5.3\%$	0%
Actual D3	0%	$73.5 \pm 18.1\%$	$21.9 \pm 18.4\%$	0%
Actual D4	0%	$18.5 \pm 22.8\%$	$73.1 \pm 24.6\%$	$6.5 \pm 16.8\%$
Actual D5	0%	$3.1 \pm 7.2\%$	$3.1 \pm 10.7\%$	$93.9 \pm 12.1\%$

Brain-to-machine-to-brain interface in a quadriplegic subject



Closed-loop BMI



Article

<https://doi.org/10.1038/s41467-023-43140-2>

Microstimulation of human somatosensory cortex evokes task-dependent, spatially patterned responses in motor cortex

Received: 30 September 2022

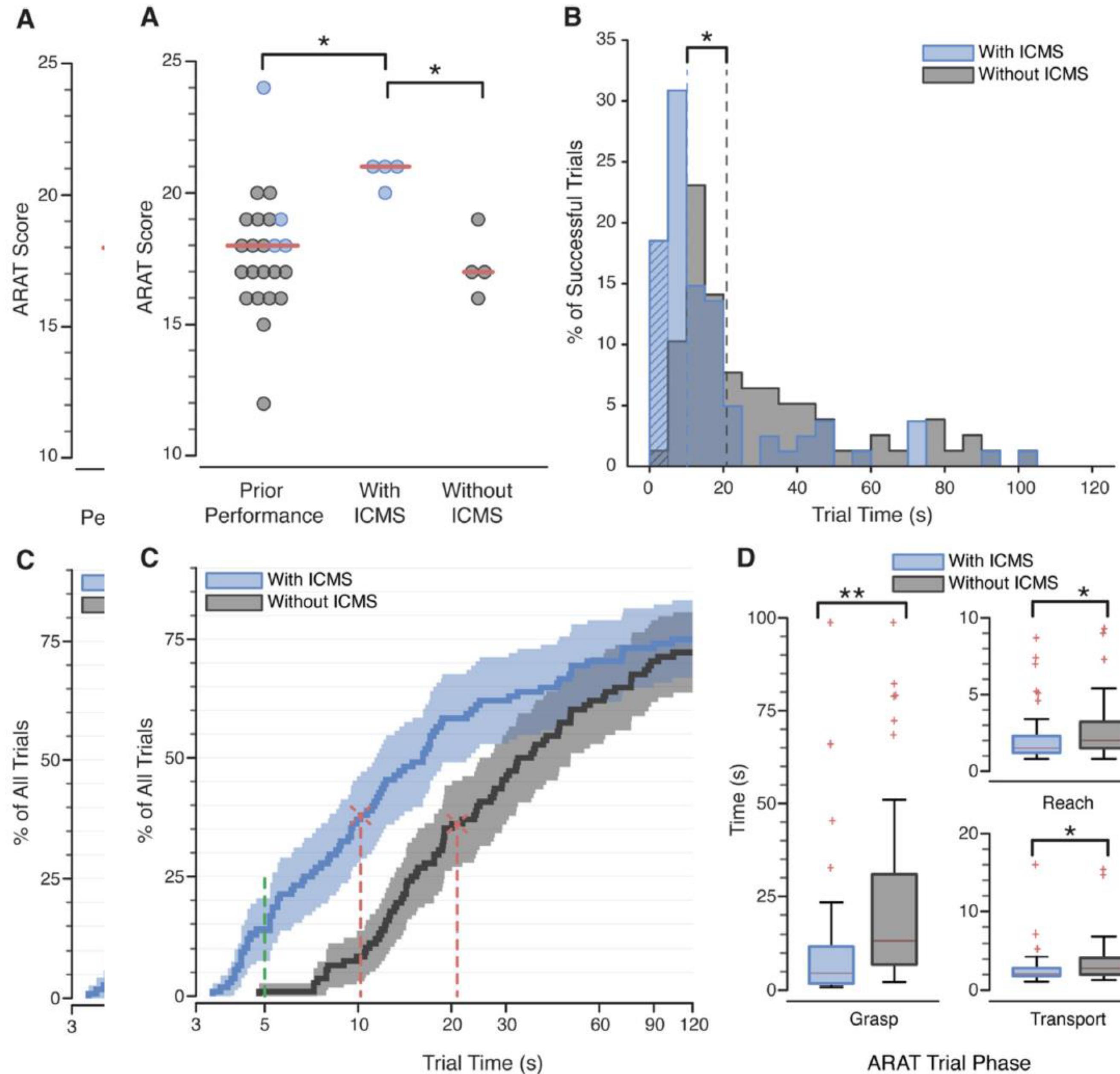
Accepted: 1 November 2023

Published online: 10 November 2023

Check for updates

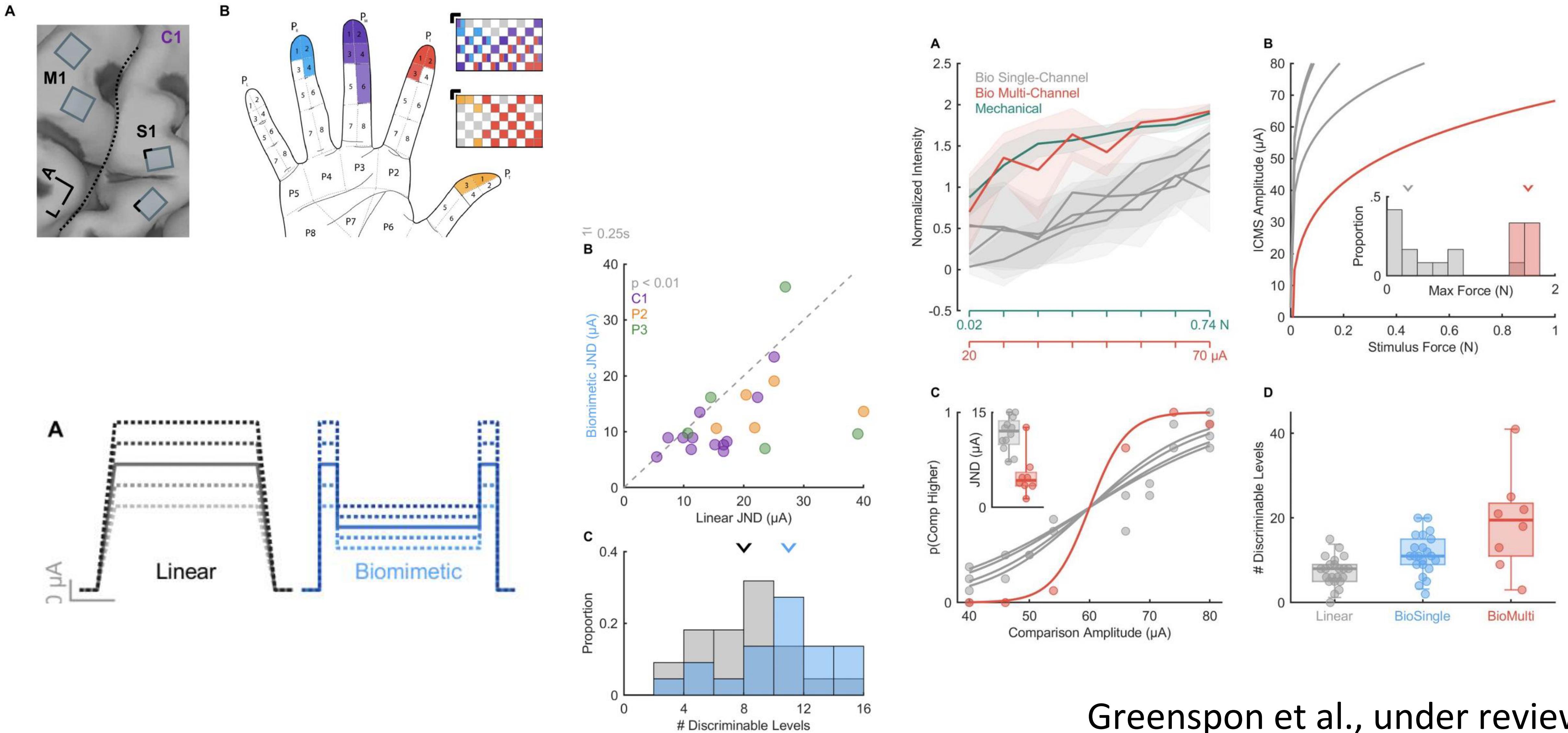
Natalya D. Shelchkova ^{1,13}, John E. Downey ^{2,13}✉, Charles M. Greenspon ^{2,13}, Elizaveta V. Okorokova ¹, Anton R. Slobinov ², Ceci Verbaarschot ^{3,4}, Qinpu He ¹, Caleb Sponheim ¹, Ariana F. Tortolani ¹, Dalton D. Moore ¹, Matthew T. Kaufman ^{1,2,5}, Ray C. Lee ⁶, David Satzer ⁷, Jorge Gonzalez-Martinez ⁸, Peter C. Warnke ^{5,7}, Lee E. Miller ⁹, Michael L. Boninger ^{3,10,11}, Robert A. Gaunt ^{3,10,11,12}, Jennifer L. Collinger ^{3,10,11,12}, Nicholas G. Hatsopoulos ^{1,2,5} & Sliman J. Bensmaia ^{1,2,5}

Closed-loop BMI



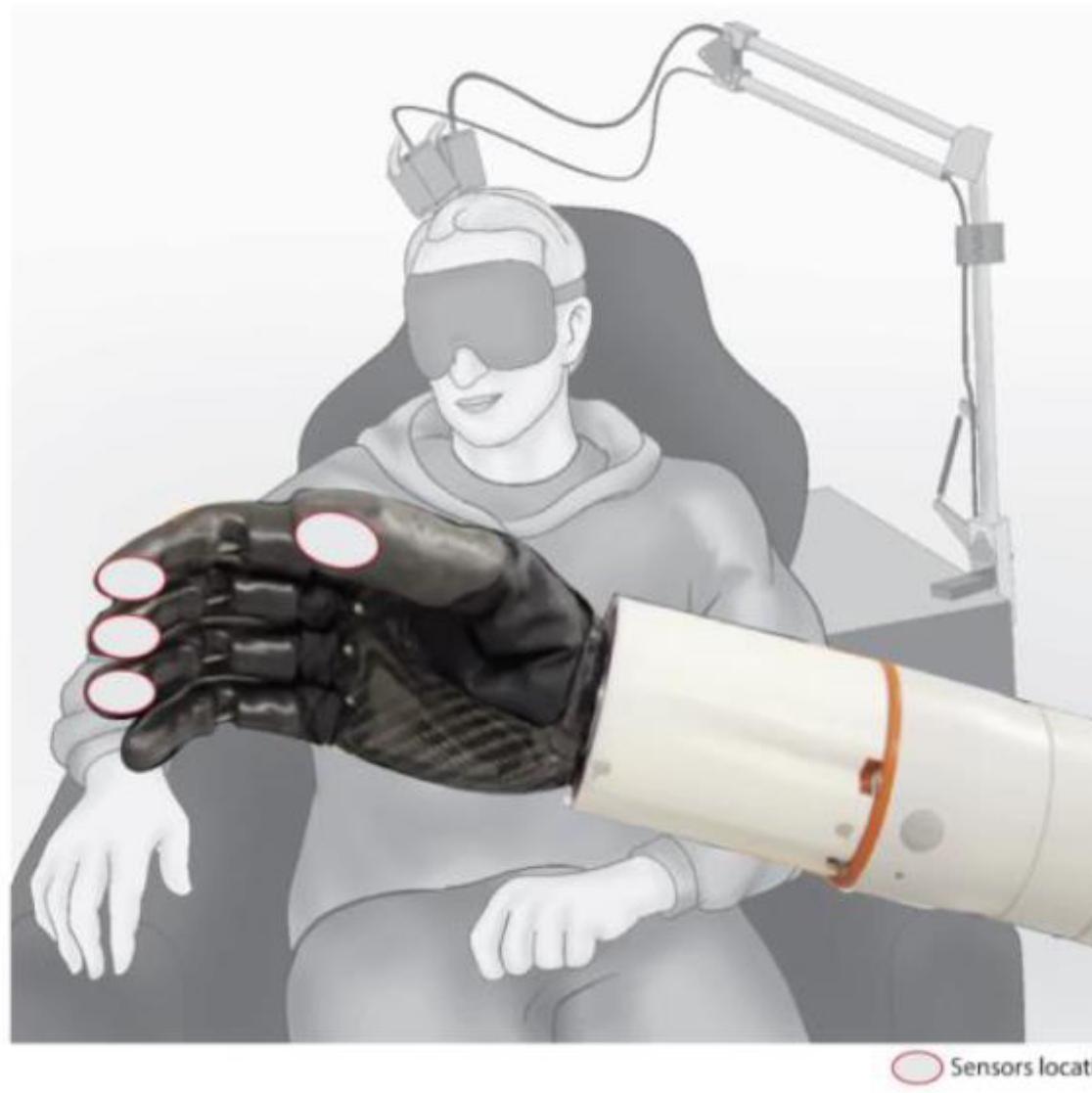
Closed-loop feedback BMI shows a significant improvement in motor performance

Biomimetic cortical stimulation

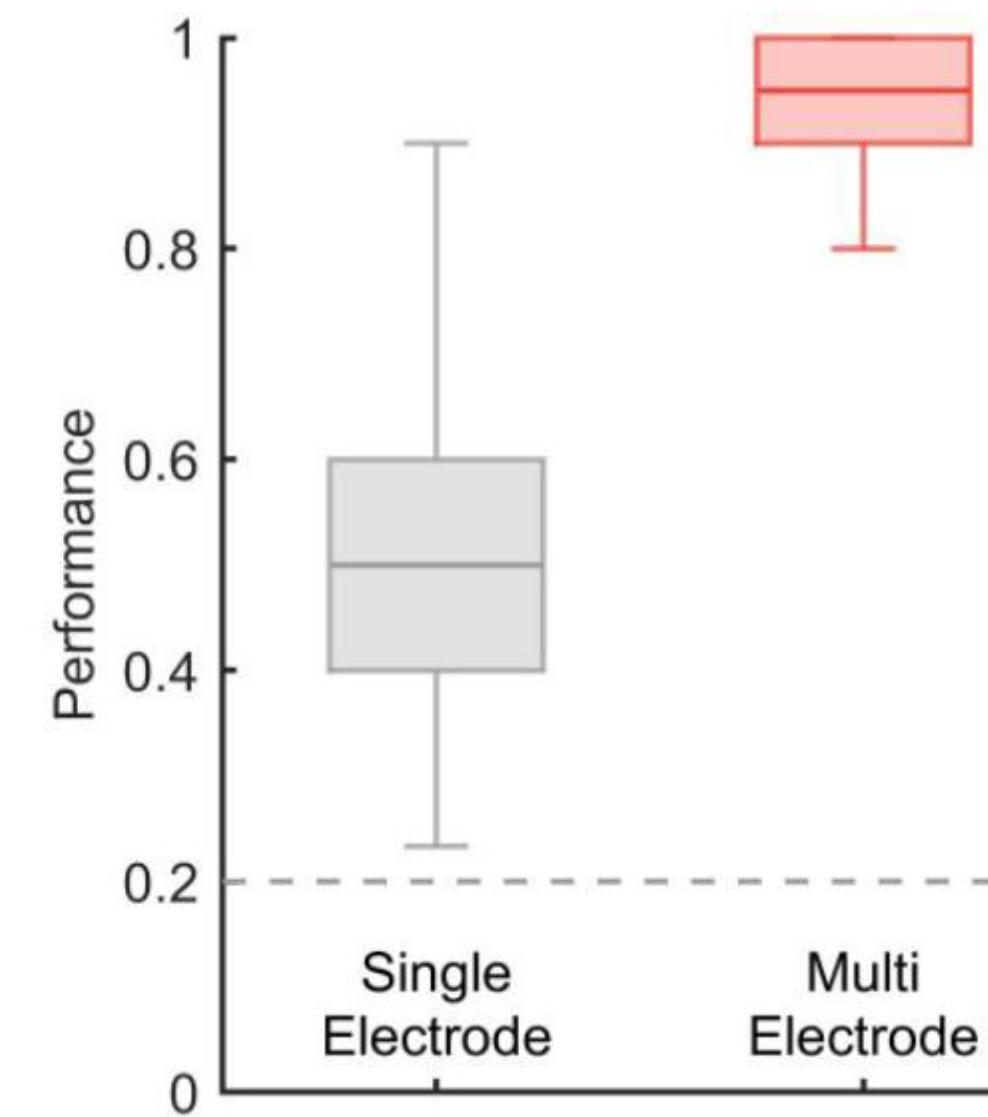


Biomimetic cortical stimulation

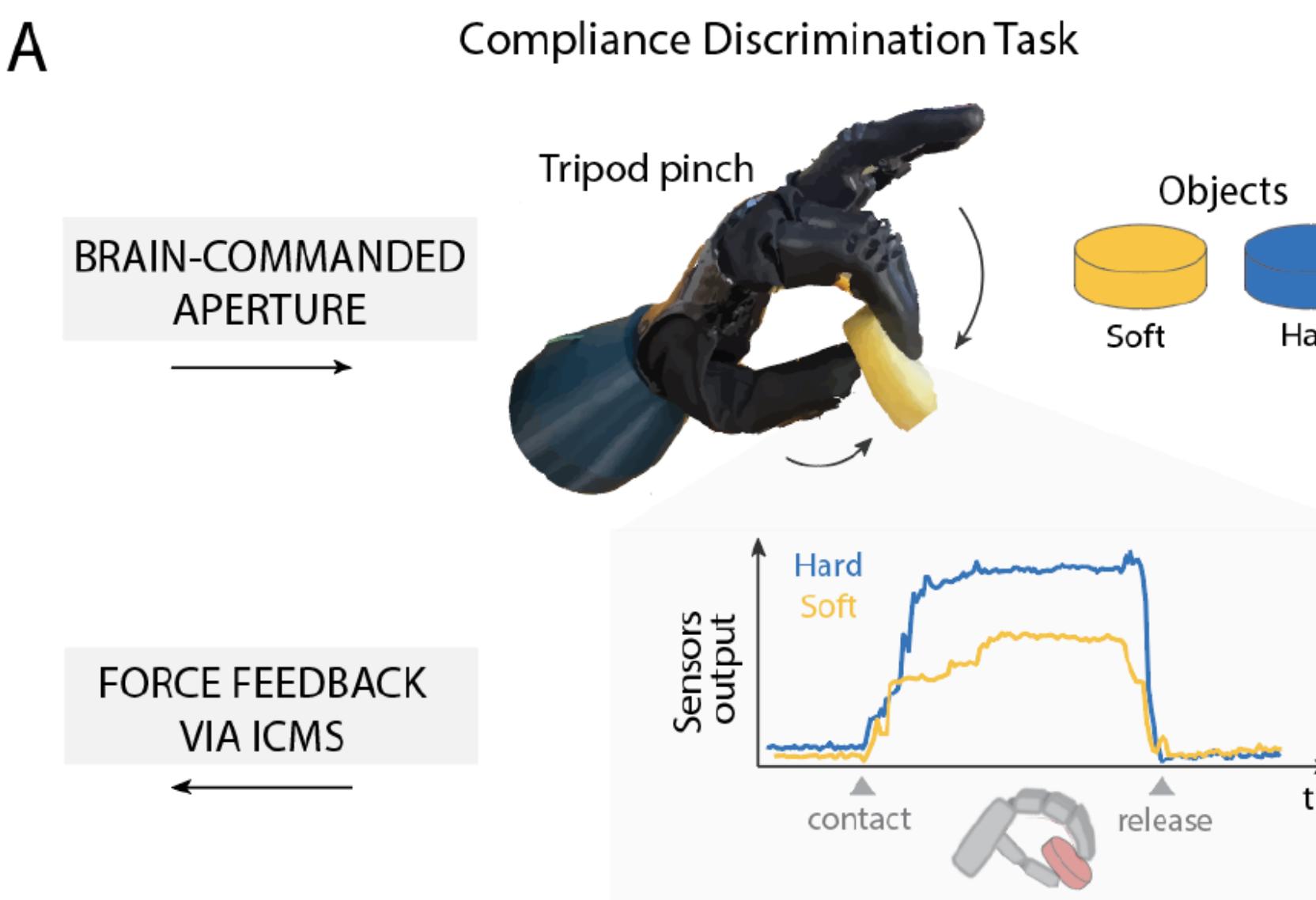
A



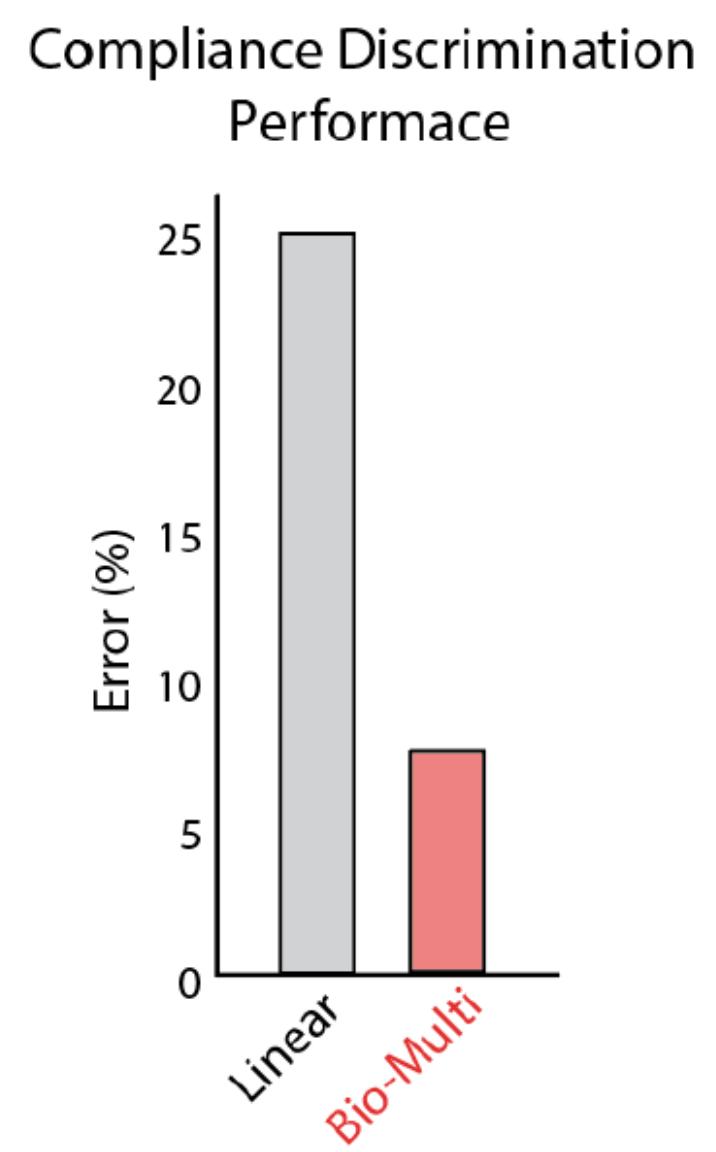
B



A



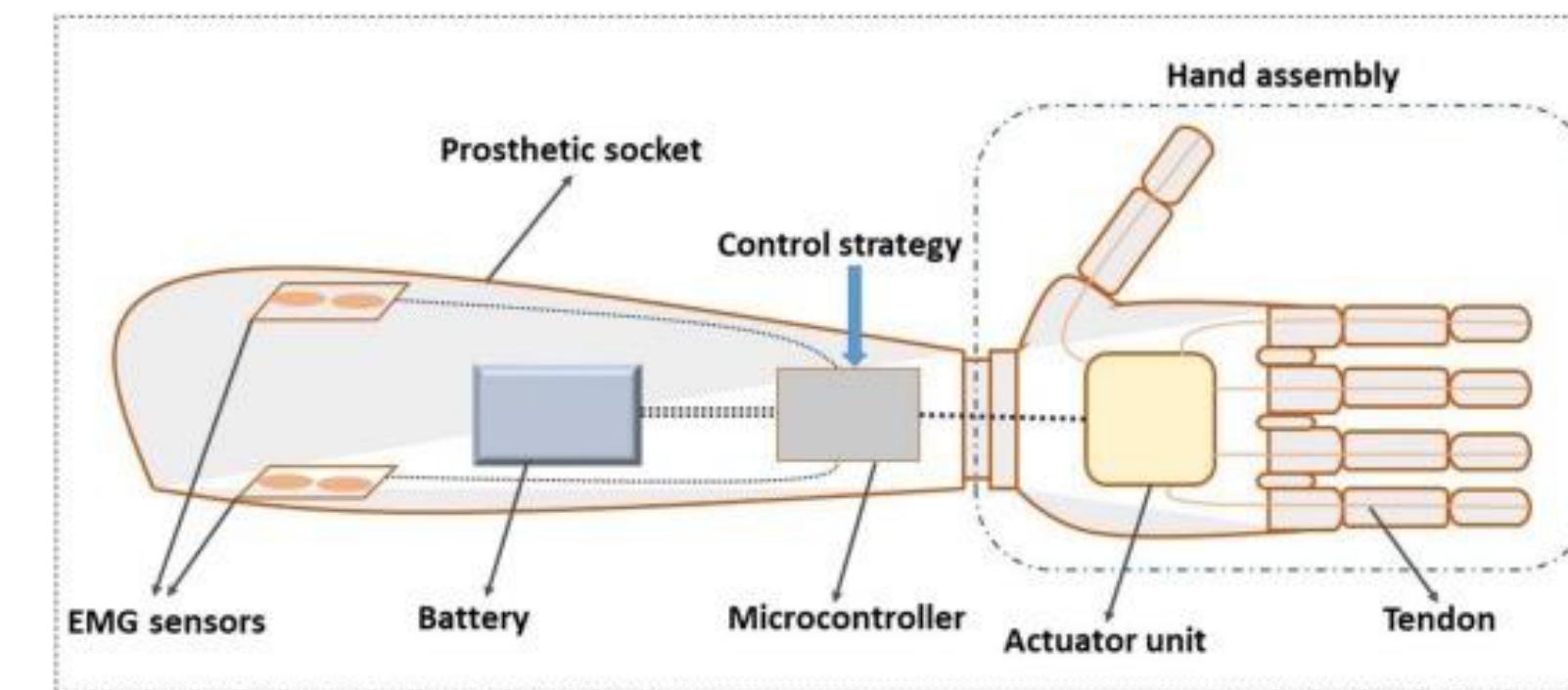
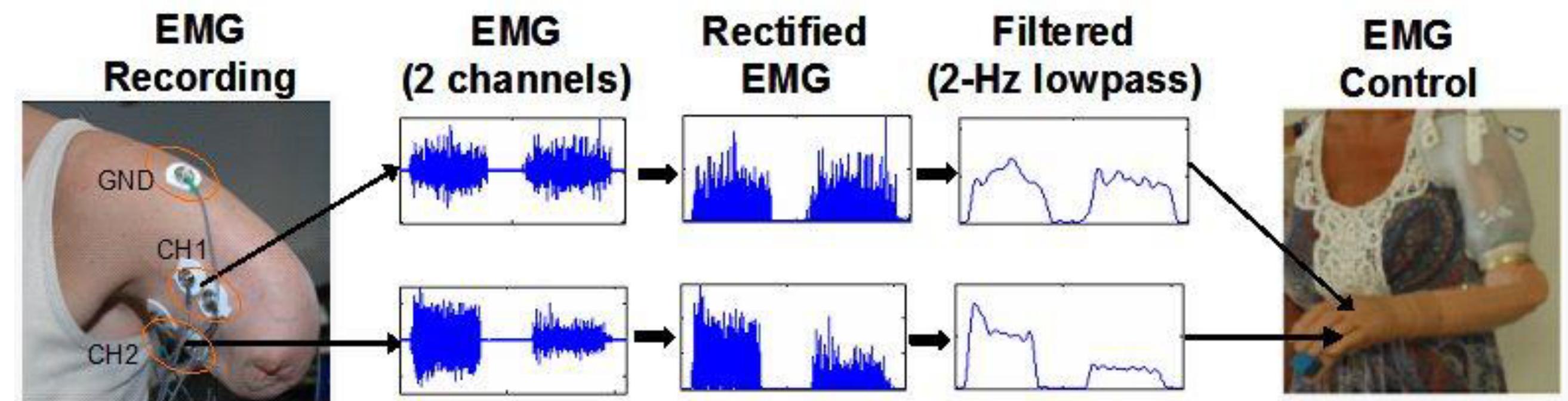
B

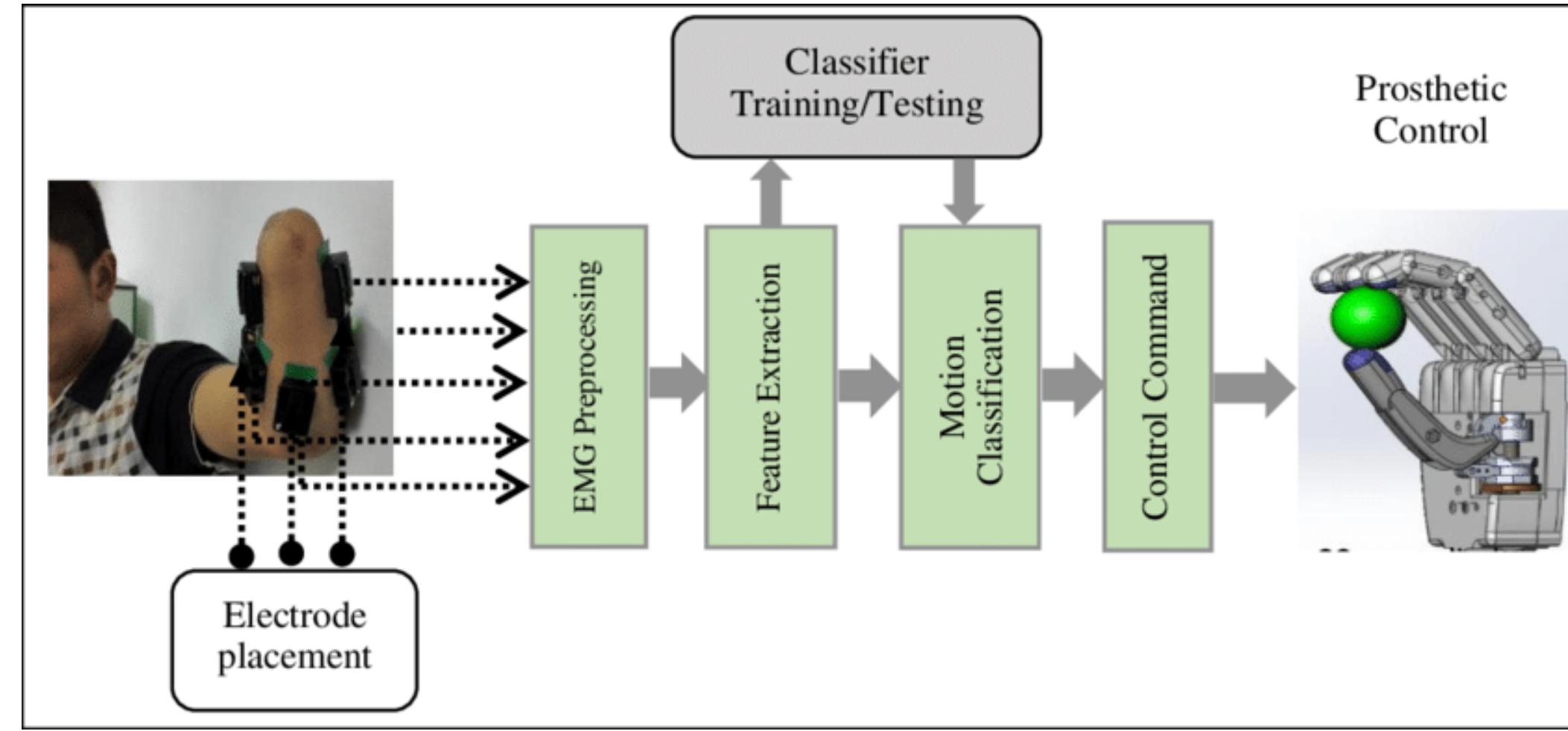


Greenspon et al., under review

Muscular control of artificial limbs

- N antagonist muscles are used to control 1 degree of freedom of the prosthesis (hand opening/closing). Often biceps/triceps or wrist extension/flexion
- An increased number of required movements makes very difficult to use this approach

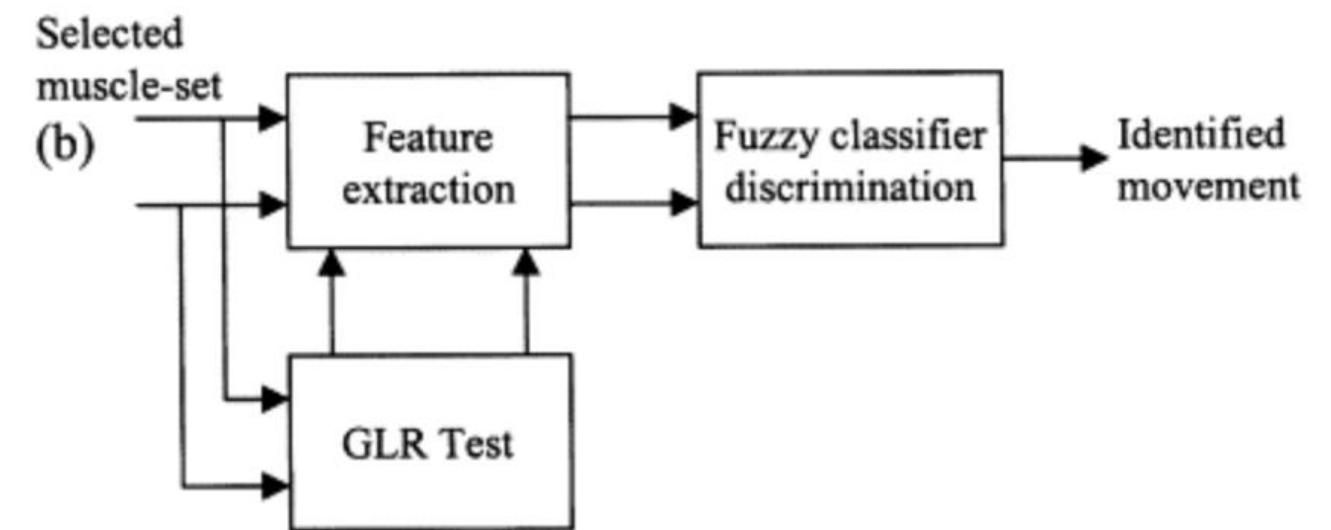
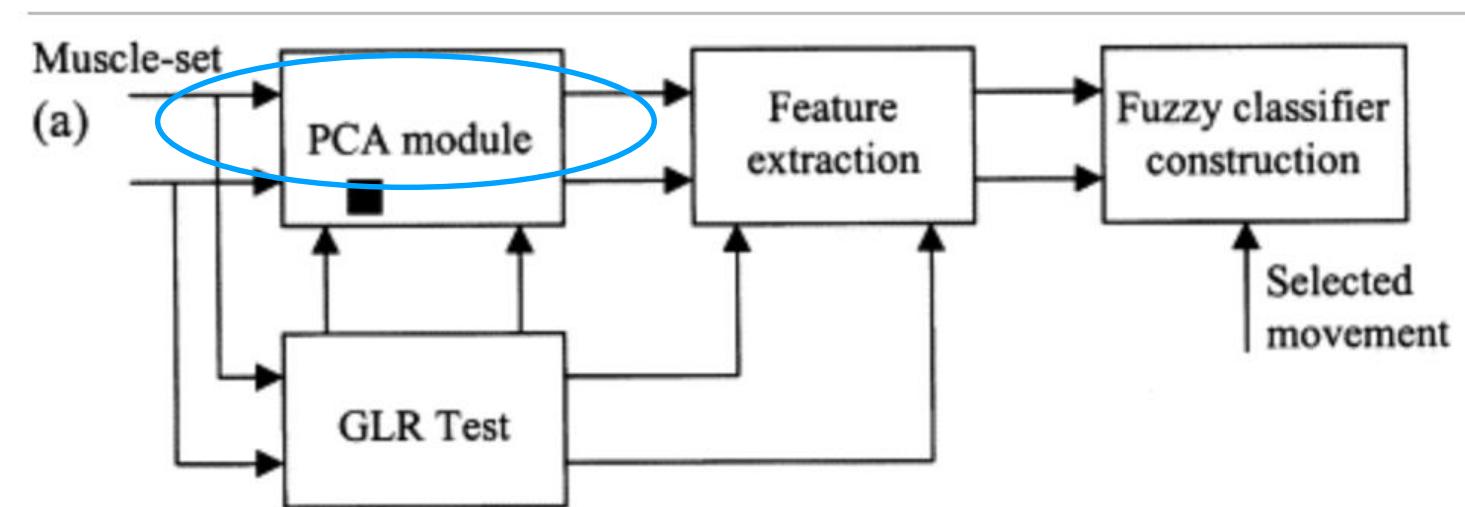




- In this case, the muscles naturally involved in the specific movement (e.g. ECR for the extension of the wrist) are no more available
- For this reason, “not- homologous” voluntary movements of the subject have to be coded as prosthesis movements (e.g. extension of the elbow for the extension of the wrist)
- This approach requires a quite long training phase and makes very difficult for the subject to easily control more than two degrees of freedom

TD Feature	Definition	References
Mean absolute value	$MAV_i = \frac{1}{N} \sum_{k=1}^N x_i(k) $	[46], [50], [52]
Integrated absolute value	$IAV_i = MAV_i * N$	[53]
Variance	$VAR_i = \frac{1}{N} \sum_{k=1}^N (x_i(k) - \bar{x}_i)^2$	[52], [54]
Mean absolute value slope	$MAVS_i = MAV_{i+1} - MAV_i$	[46]
Willison amplitude	$WAMP_i = \sum_{k=1}^N f(x_i(k) - x_i(k+1))$ with $f(x) = 1 \text{ if } x > x_{th}, 0 \text{ otherwise}$	[54]
Zero crossing	$ZC_i = \sum_{k=1}^N f(k)$ with $f(k) = 1 \text{ if } x_i(k) * x_i(k+1) < 0 \text{ and } x_i(k) - x_i(k+1) > x_{th}$	[46]
Slope sign changes	$SSC_i = \sum_{k=2}^{N-1} f[(x_i(k) - x_i(k-1)) * (x_i(k) - x_i(k+1))]$ with $f(x) = 1 \text{ if } x > x_{th}, 0 \text{ otherwise}$	[55]
Waveform length	$WL_i = \sum_{k=1}^{N-1} (x_i(k) - x_i(k+1))$	[46], [55]
TSD Feature	Definition	References
Autoregressive coefficients	$x_i(k) = \sum_{j=1}^N a_j x_i(k-j)$, n^{th} order AR model	[53], [56]–[58]
Cepstral coefficients	$c_1 = -a_1; c_i = -a_i - \sum_{k=1}^{i-1} (1 - \frac{k}{i}) a_n c_{i-k}$ $1 \leq k \leq n$ and a_i are the AR coefficients	[52]
FD Feature	Definition	References
Mean of signal frequencies	$FMN_i = \sum_{j=1}^M (f_j p_j) / \sum_{j=1}^M (p_j)$	[7]
Frequency ratio	$FR_i = \frac{\min(FFT(x_i))}{\max(FFT(x_i))}$	[7]
TSC or TF Feature	Definition	References
Short-time Fourier transform	$STFT[k, m] = \sum_{r=1}^{N-1} x[r] g[r-k] e^{-j2\pi m i/N}$ where g , k , and m are the window function, the time sample, and frequency bins, respectively.	[7]
Wavelet transform	Continuous WT (CWT) produces a good frequency resolution Δf in long time windows (low frequencies) and a good time localization Δt at high frequencies $CWT_x(\tau, a) = \frac{1}{\sqrt{a}} \int x(t) \Psi(\frac{t-\tau}{a}) dt$ where t and a are the translation and scale parameters and Ψ is the mother wavelet function	[47], [51]
Wavelet packet transform	WPT is a generalized version of the continuous and discrete WT.	[47], [51], [59]

PCA or similar could be necessary to select few more information muscles

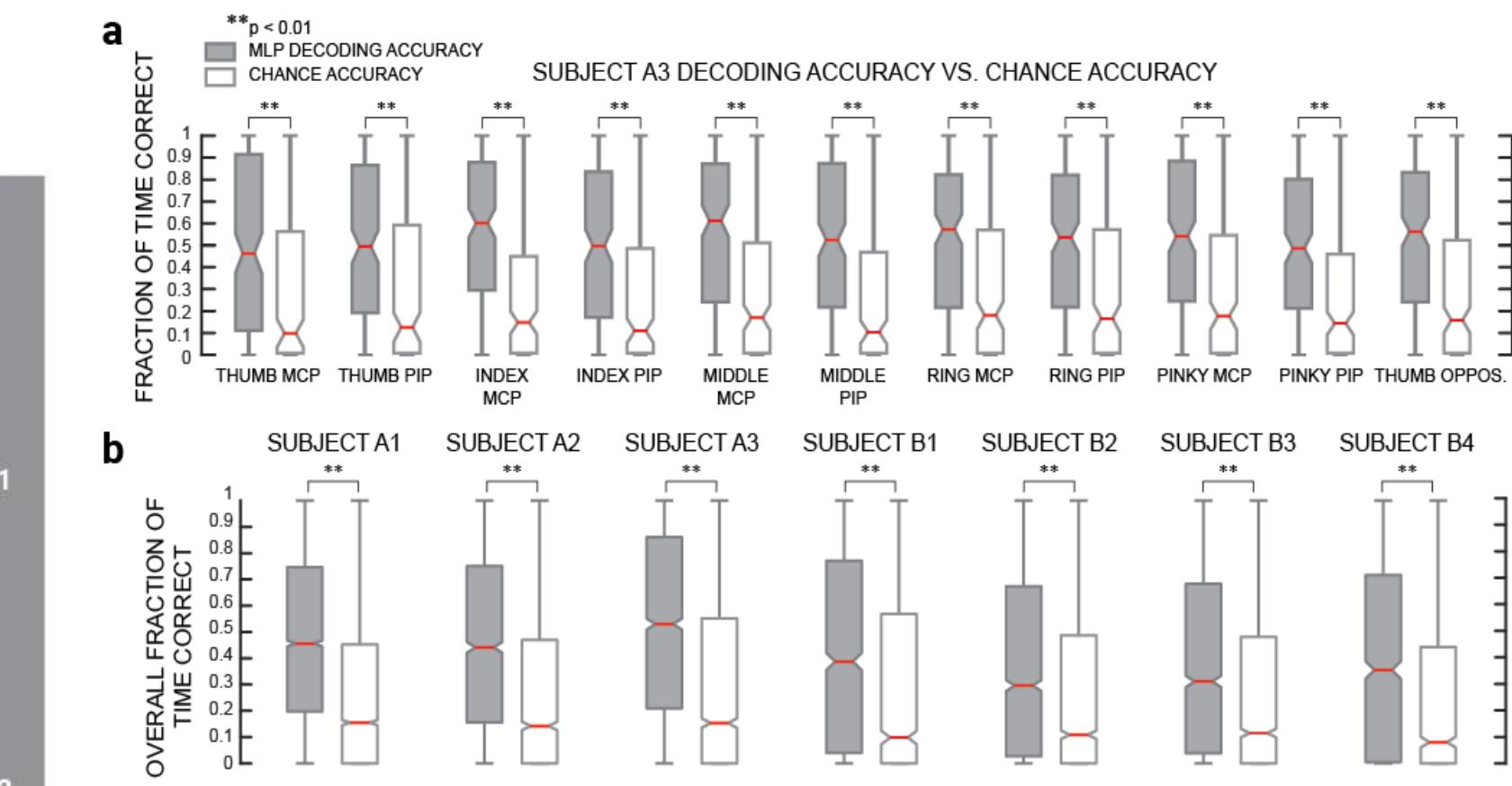
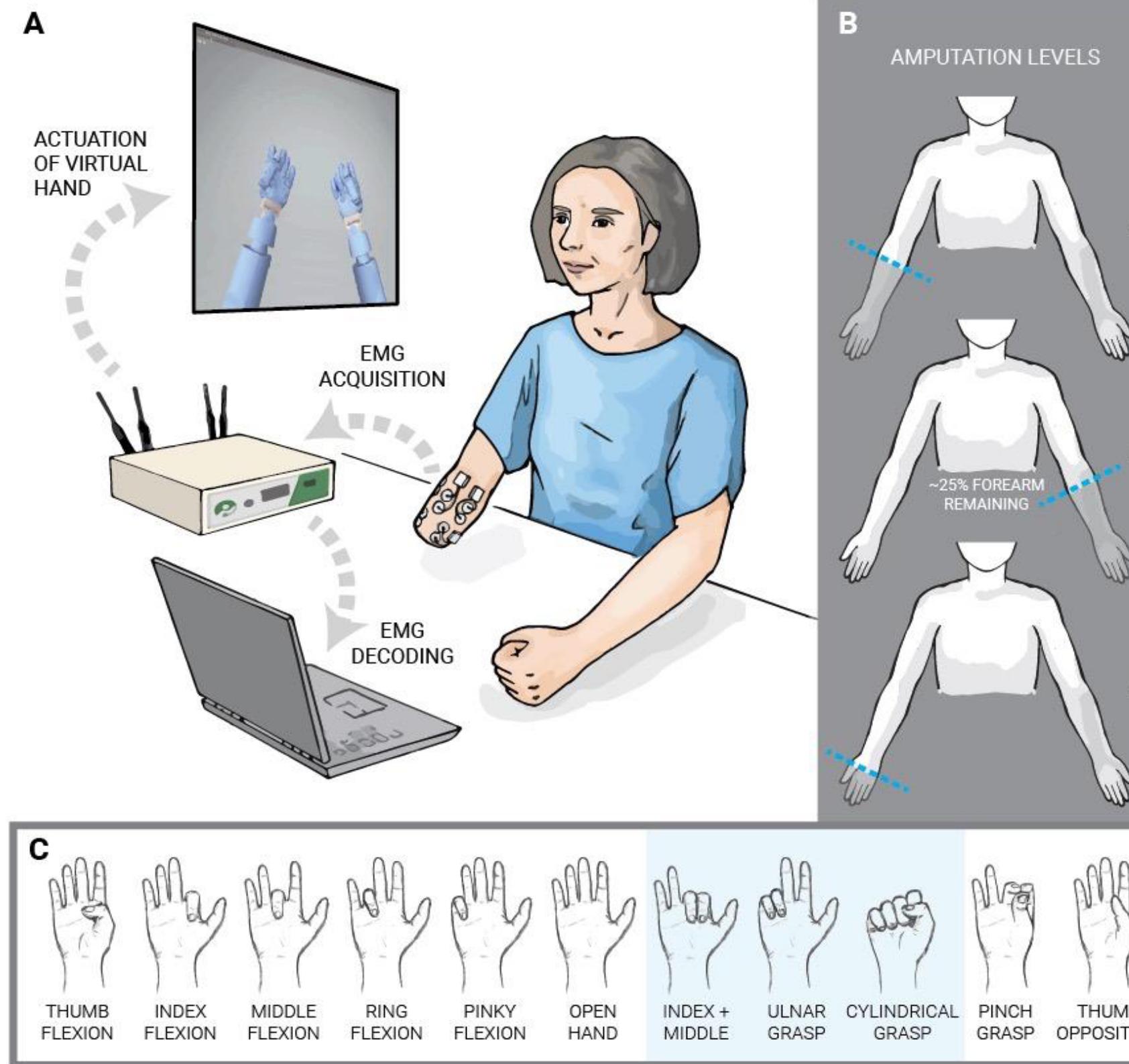


- Supervised learning classifier to link EMG signals (features) to desired hand movements
- More or less anything has been tried (including majority voting)
- Make a fair comparison!!

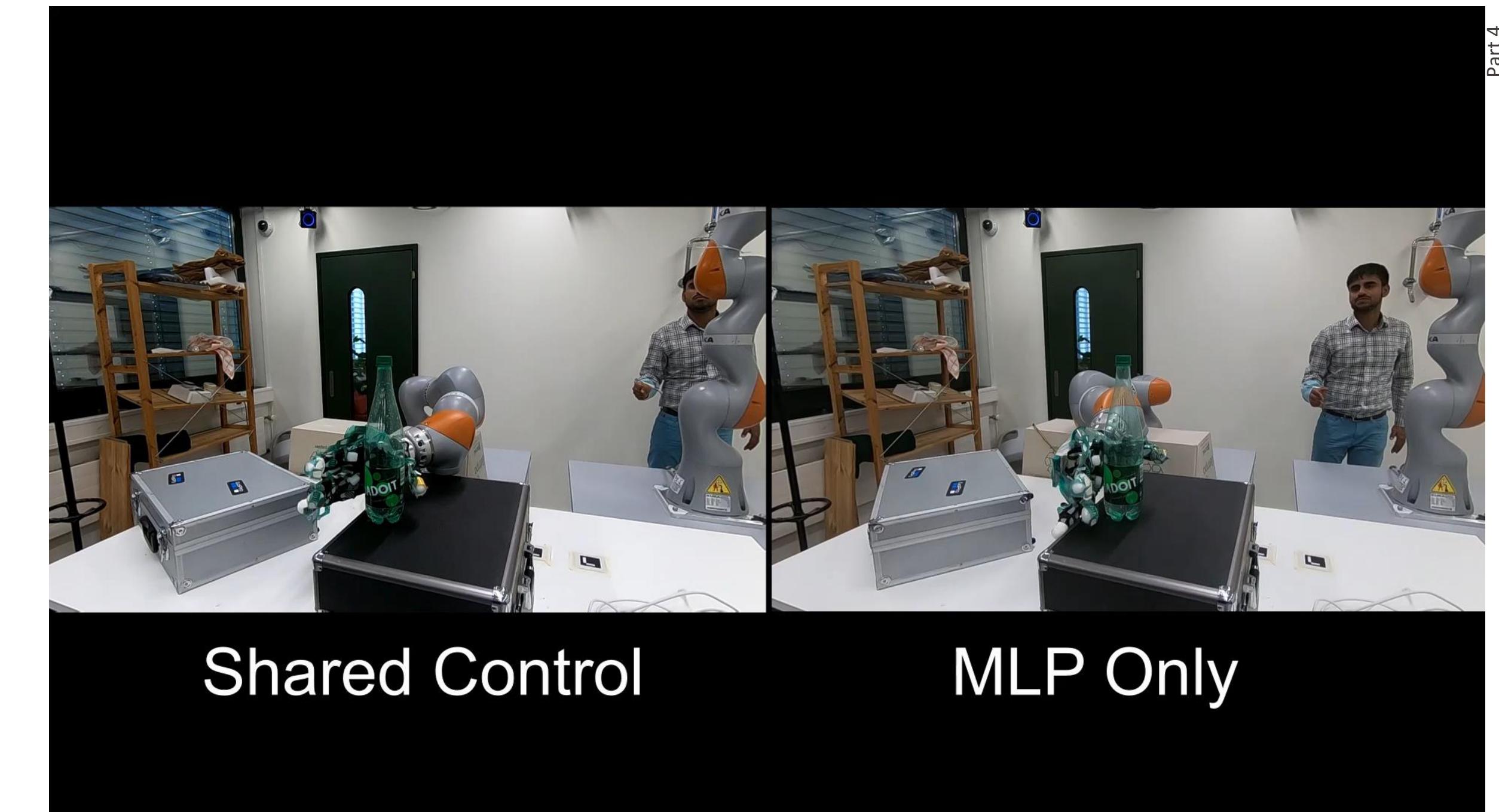
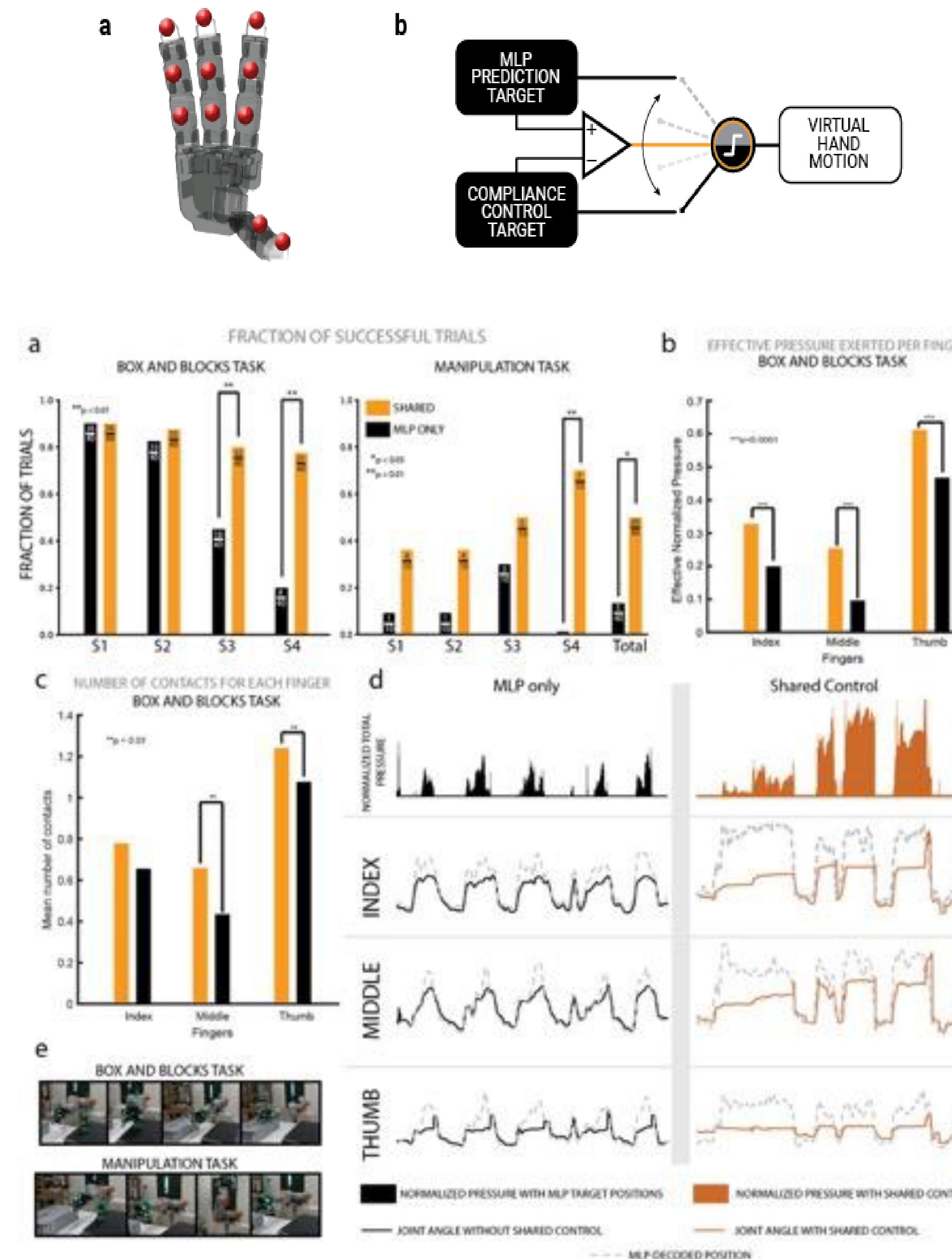
TABLE IV
SOME PATTERN RECOGNITION BASED CONTROL OF UPPER LIMB PROSTHESIS.
A = amputee subjects, H = healthy subjects, LD = limb deficiency subjects

Classifier	EMG channels	Classes	Features	Subjects involved	References
MLP	2	4	MAV, MAVS, ZC, SSC, WL	9H+6A	Hudgins et al [46]
Fuzzy	2	6	IAV, VAR, AR, CC, adaptive CC	6H	Park and Lee [52]
LDA, MLP	2	4	-	16H	Englehart et al [51]
Fuzzy	2	4	MAV, MAVS, ZC, WL	4H	Chan et al [65]
PCA, LDA	2,4	4,6	-	11H	Englehart et al [47]
PCA, LDA	4	6	STFT, WT, WPT	12H	Englehart et al [12]
-	3,4	3,4	Fuzzy	3H+1A+1LD	Ajiboye and Weir [67]
HMM, MLP	4	6	-	12H	Chan and Englehart [66]
GMM, LDA, MLP	4	6	TD, RMS, AR	12H	Huang et al [68]
LDA, MLP	4	8	WPT	10H	Chu et al [64]
SVm, GDA	3	8	AR, histogram	1H+2A	Liu et al [70]
SVM, LDA, MLP	4	5	single and multi TD/FD	11H	Oskoei and Hu [49]
SVM	7	8	RMS	3H	Shenoy et al [71]
HMM, bayes	4	9	-	10H	Chu and Lee [69]
MLP	12/32	12	MAV, VAR, WL, W	5H+1A	Tenore et al [54]
LDA	12	10	MAV, ZC, WL, SSC	5A	Li et al [72]

EPFL EMG control – A little help?



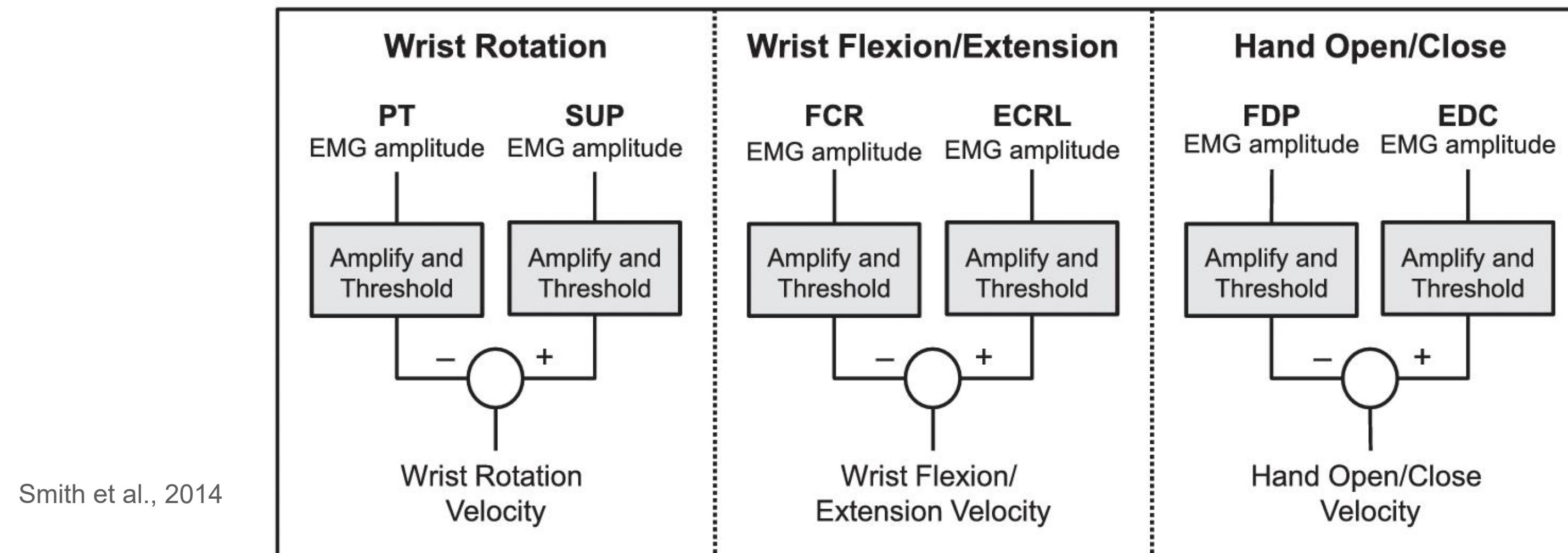
- Single finger decoding using EMG signals
- One implanted patient from UCSC - Loretana
- Two patients from collaboration with hospitals Chuv (Lausanne, CH) and Villa Beretta (Lecco, Italy)



When a hand is not contacting an object, the user controls the robotic hand with the output of EMG decoding

When the hand makes contact with an object, the compliance controller automates hand conformation around the object, allowing a high degree of grasp stability

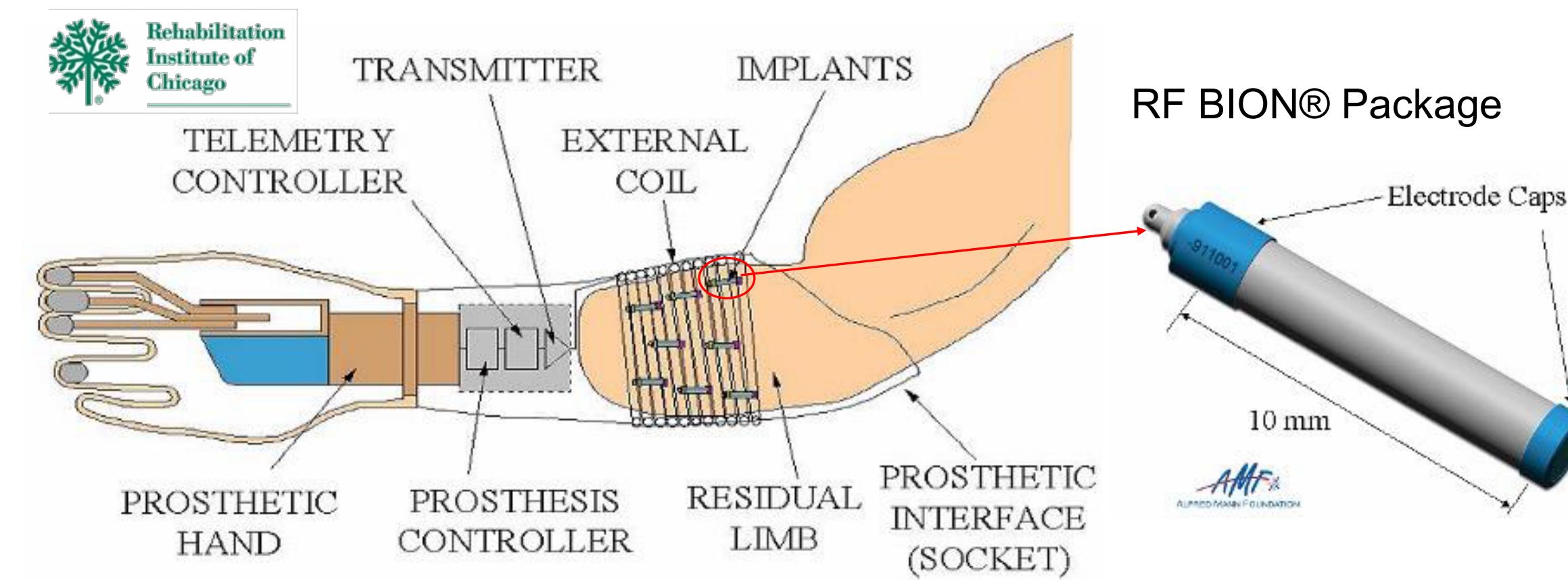
- Clinically available myoelectric control strategies do not allow simultaneous movement of multiple degrees of freedom (DOFs)
- The use of implantable devices that record intramuscular EMG signals could overcome this constraint
- Intramuscular EMG signals can be recorded using percutaneous fine wire electrodes inserted using needles
- The use of iEMG can allow to use proportional control (but of course also pattern recognition)

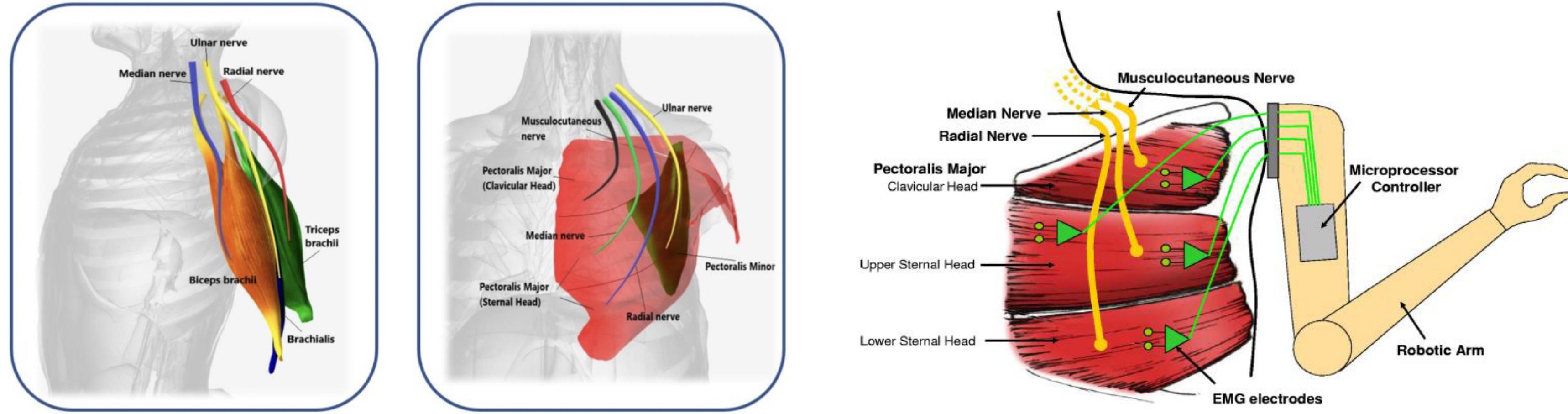


- Sense myoelectric signal at its source, so it acts as an amplifier of the neural command.
- Use inductive coupling to pass power into devices and signal out of device w/o breaking the skin

Multifunction Prosthesis Control Using Implanted MyoElectric Sensors (IMES)

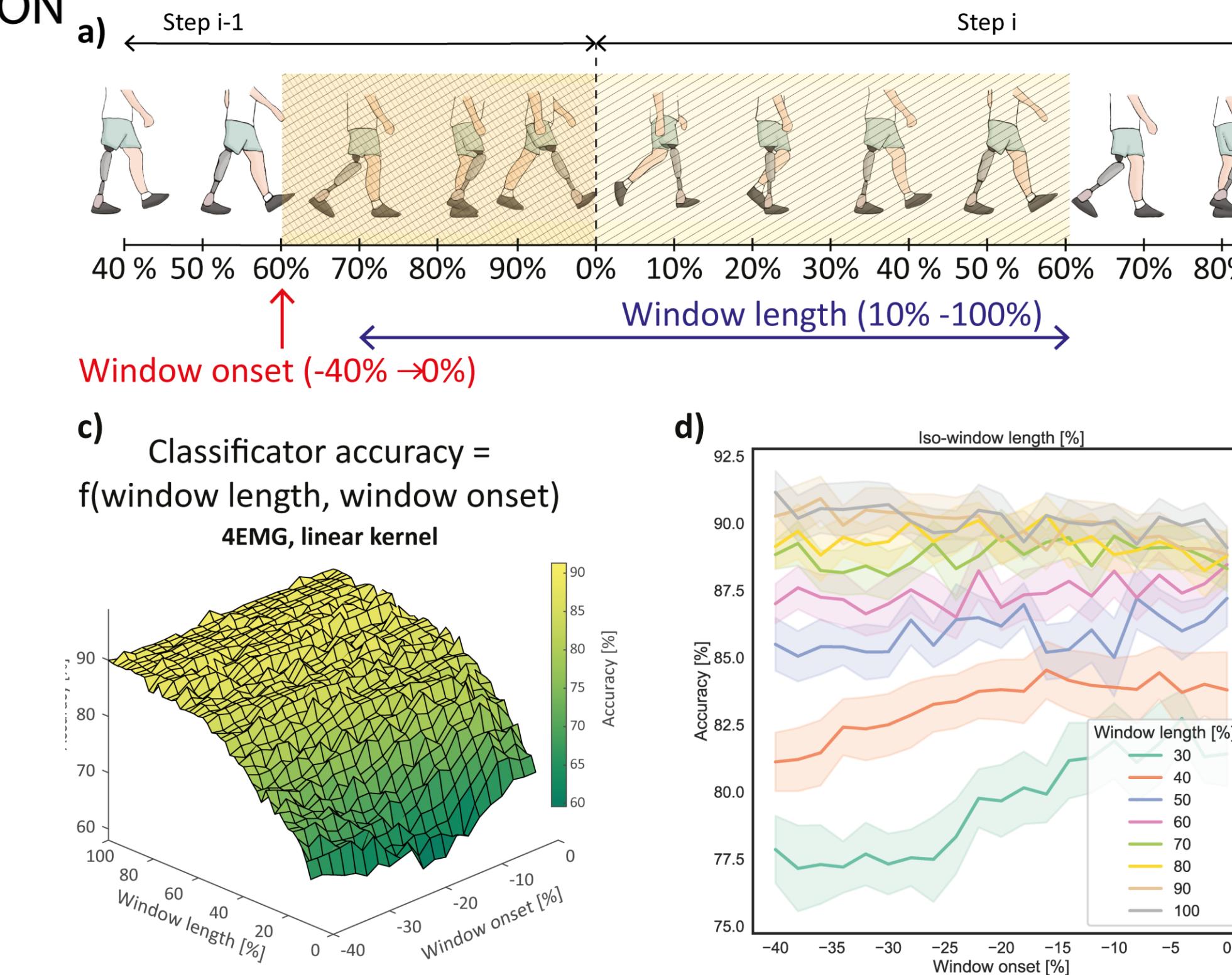
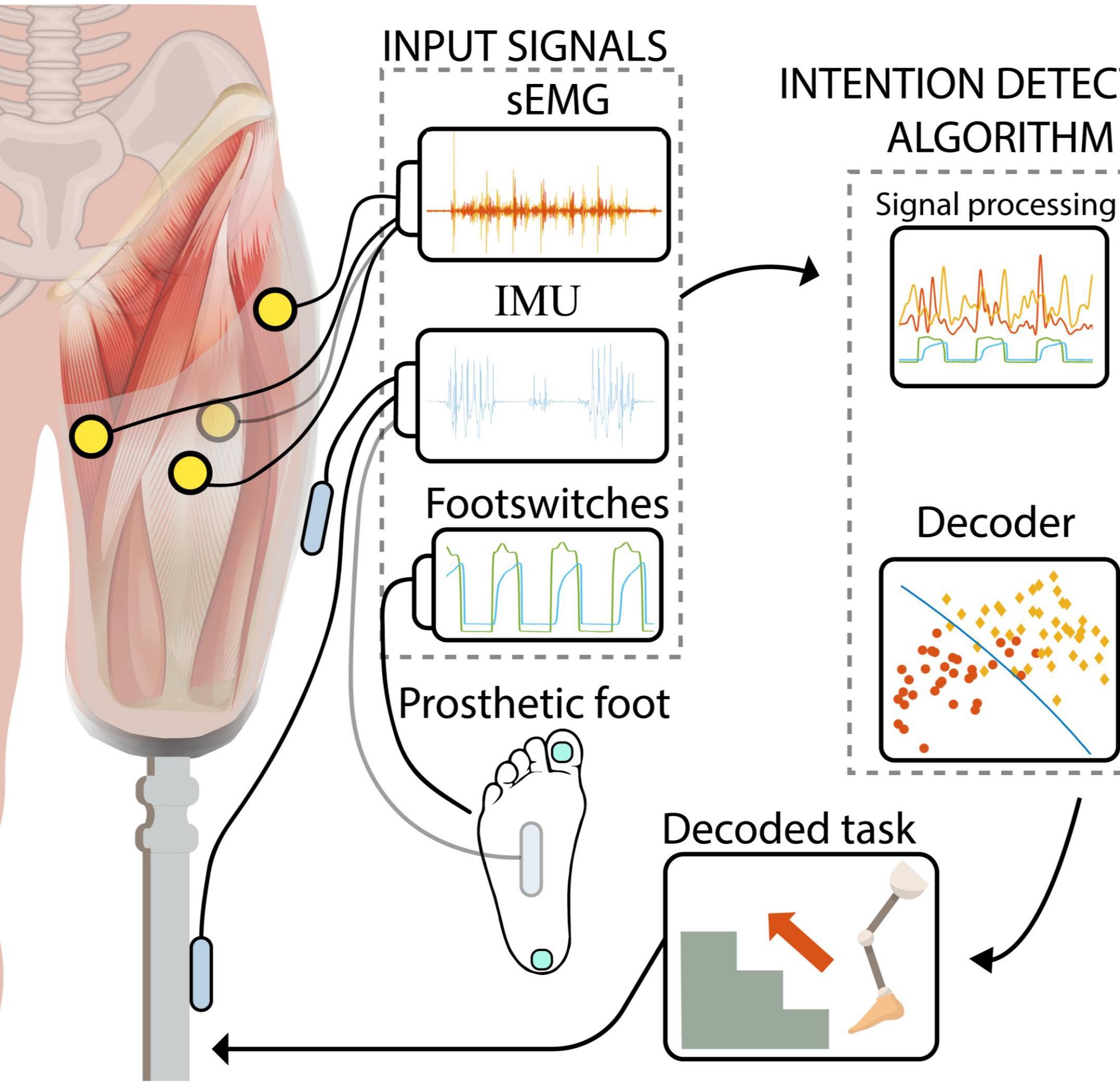
Smith et al., 2014





- A surgical technique called targeted muscle reinnervation (TMR) transfers residual arm nerves to alternative muscle sites
- After reinnervation, these target muscles produce electromyogram (EMG) signals on the surface of the skin that can be measured and used to control prosthetic arms

Bidirectional neurocontrolled leg prostheses



Conclusions

- Artificial limbs can be bidirectionally controlled in several ways
 - Non-invasive interfaces for decoding (EMG, EEG) and encoding (vibrators, transcutaneous electrical stimulation)
 - Invasive interfaces for decoding and encoding (ECoG, intracortical, peripheral implants)
- The choice must be done taking into account the residual skills of the subjects AND their preferences
- The different neurotechnological “tools” must be integrated accordingly

